

Final Year Project Report

Full Unit - Interim Report

Comparison of Ridge Regression Algorithm against Others When Solving Boston Housing Problem

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Declaration

This report has been prepared on the basis of my own work. Where other published and unpublished source materials have been used, these have been acknowledged.

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Abstract

For most people buying a house will be one of their most important and expensive economic decisions that they will take in their lives.[1] Because of this it would be logical to say that being able to accurately predict the prices of said houses would be of extreme value to people. One possible way to make these predictions would be to create a machine learning model that, given a certain amount of features from each house, would be able to create an accurate prediction of their price.[2] There are a wide range of different machine learning algorithms that could be used to solve this problem. However, this project will focus on two: Ridge Regression and Decision Trees.

In this project I plan on implementing both of these algorithms and comparing their performance on one another to see which one is more effective at resolving the Boston housing problem. To measure their accuracy I will use two metrics specifically used for regression problems, mean absolute error (MAE) and root mean squared error (RMSE). MAE returns the average residual of the predictions that each model makes, this is useful as it can be directly compared against the other model for differences.[3] However, RMSE returns the square root of the average residual of the predictions. This highlights larger errors caused by the model which is practical as it helps to differentiate the performance of both algorithms.[4]

The dataset that will be used to judge the different algorithms is called the Boston housing problem datasets. The dataset is comprised of 506 entries, each having a total of 14 features which describe a property inside of the Boston Massachusetts area.[5] Two variations of this dataset will be used, one with full 14 features and another with a lower number 5. Two models will be created, one with each algorithm and each of these models will be trained and tested on both of these datasets. Their performance will then be analysed, compared, and finally conclusions will be drawn from the results in order to define the effectiveness both algorithms.

Project Specification

Your project specification goes here.

Chapter 1: Introduction

Since humans began roaming the earth there have always been three main things that we have needed to survive: food, water, and shelter. In modern developed civilisations food and water has become relatively easy to come by with affordable versions of both being available to most people. However, shelter seems to only become more expensive with time while income stays the same. This can be seen in the UK where the median price of residential houses increased by 14% while income decreased by 1% from 2020 to 2021, making the issue of housing one of the most prevalent ones in recent times.[6] Also, With the advent of the internet and its capability of delivering large scale of information to users, the options of houses available are in-numerous. Because of this it can be an extremely overwhelming task to look for a suitable place to live that both fits the budget but is also reasonably priced considering the aspects and features of the house. House prices also play a significant role in the economy, one of these is being linked to consumer spending. If a homeowner knows that the value of his home increases then he feels confident and is likely to spend more in goods or services or to pay off their debts. If they know the value has decreased then the opposite occurs, homeowners become less confident meaning they are likely to spend less and save more.[7] Taking all of these factors into consideration it can be said that having an accurate method of estimating house prices is of importance from an individual looking for a home all the way to governments trying to predict what economic measures they should take next.

The first step in order to generate accurate estimates is to gather and analyse data on past houses which have been sold and try to find a pattern regarding their value. Specifically between the price they were sold for and features that describe the houses, such as the size of the house, area it is located in, and so forth. This analysis used to be done through traditional statistical methods, however, recently a new approach has emerged called machine learning. Here a machine is responsible for processing and learning from the data it is given in order to make predictions on new data autonomously.[8] When it comes to predicting house prices the machine learning model is given data consisting of past sales of houses where each house sold is also accompanied with a number of features that describe it as mentioned before. The system can then learn how these features influenced the price of the houses and use this information to make predictions on new houses based only on its corresponding features.[9]

Chapter 2: Theoretical Background

Within machine learning there are two main methodologies utilised when trying to solve a problem, supervised and unsupervised learning. Unsupervised learning is used when the data being analysed does not need to be labelled, but instead needs to be sorted into groups by their features. On the other hand, supervised learning is concerned with providing labels to unlabelled data in an dataset, such as a list of houses without a price attached to them.[10] Within supervised learning there is once again two different types of problems that occur: classification and regression problems. Classification problems occur when the list of possible classifications is finite such as identifying a handwritten digit as its correct number. However, in the Boston housing problem the list of possible labels is infinite as they could be any price, i.e., any real number.[11] This is called a regression problem and will be the focus of the research paper.

2.1 Simple Multiple Linear Regression

Regression as a type of machine learning problem comes from the statistical method of regression, which has the goal of determining the relationship between a dependent and independent variable(s).[12] This is then expanded to create regression models which can be formally defined as using an independent variable ' x ' to predict a dependent variable ' y '. [13] With the Boston housing problem the independent variable x are the features of each house such as, per capita crime rates per town, while the dependent variable y are the prices of the houses.[14]

There are many variations of the regression algorithm that can be used in a machine learning model, however simple linear regression is arguably the most simplistic ones and serves as a base for the others. The equation used in a simple linear regression model has the form:

$$y = \beta_0 + \beta_1 x + \epsilon$$

where y and x are the independent and dependent variable respectively. β_0 is the intercept of the line created by the equation with the y -axis, this is known as the constant term as it does not change. β_1 is the regression coefficient which defines the slope of the line. And lastly ϵ is the random error term or the random deviation which represents the difference between the predicted value and the real-life value.[15, 16] The goal of this machine learning method is to fit the best possible line to the given data, which is achieved by finding the optimal weights of the regression coefficients β . For this a cost function is used which calculates the difference between the predicted value and the expected values of every sample in the training dataset and returns it as a single real number.[17] This cost is then minimised by modifying the regression weights β , when the cost function cannot be minimised further the most optimal coefficients have been achieved.[18]

2.2 Ridge Regression

2.3 Random Forest

Chapter 3: **Methodology**

Chapter 4: **Headings**

Your report will be structured as a collection of numbered sections at different levels of detail. For example, the heading to this section is a first-level heading and has been defined with a particular set of font and spacing characteristics. At the start of a new section, you need to select the appropriate L^AT_EX command, `\chapter` in this case.

4.1 Second Level Headings

Second level headings, like this one, are created by using the command `\section`.

4.1.1 Third Level Headings

The heading for this subsection is a third level heading, which is obtained by using command `\subsection`. In general, it is unlikely that fourth or fifth level headings will be required in your final report. Indeed it is more likely that if you do find yourself needing them, then your document structure is probably not ideal. So, try to stick to these three levels.

4.2 A Word on Numbering

You will notice that the main section headings in this document are all numbered in a hierarchical fashion. You don't have to worry about the numbering. It is all automatic as it has been built into the heading styles. Each time you create a new heading by selecting the appropriate style, the correct number will be assigned.

Chapter 5: Presentation Issues

5.1 Figures, Charts and Tables

Most final reports will contain a mixture of figures and charts along with the main body of text. The figure caption should appear directly after the figure as seen in Figure 5.1 whereas a table caption should appear directly above the table. Figures, charts and tables should always be centered horizontally.



Figure 5.1: Logo of RHUL.

5.2 Source Code

If you wish to print a short excerpt of your source code, ensure that you are using a fixed-width sans-serif font such as the Courier font. By using the `verbatim` environment your code will be properly indented and will appear as follows:

```
static public void main(String[] args) {  
    try {  
        UIManager.setLookAndFeel(UIManager.getSystemLookAndFeelClassName());  
    }  
    catch(Exception e) {  
        e.printStackTrace();  
    }  
    new WelcomeApp();  
}
```

Chapter 6: **References**

Use one consistent system for citing works in the body of your report. Several such systems are in common use in textbooks and in conference and journal papers. Ensure that any works you cite are listed in the references section, and vice versa.

Chapter 7: **Project Information and Rules**

The details about how your project will be assessed, as well as the rules you must follow for this final project report, are detailed in the project booklet.

You must read that document and strictly follow it.

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