Implementing of RankNet

Ryan Greenup

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1 Introduction

Ranknet is an approach to *Machine-Learned Ranking* developed by Microsoft as early as 2004 [2], although previous work in this area had already been undertaken as early as the 90s (see generally [4, 3, 4, 5, 11]) these earlier models didn't perform well compared to more modern machine learning techniques [7, §15.5].

Machine-Learned Ranking is the process of applying machine learning algorithms to ranking problems, it often referred to as *Learning to Rank*/ [6].

2 Motivation

Much data cannot clearly categorised or quantified even if it there is a capacity to compare different samples, the motivating example is a collection of documents, it might be immediately clear to the reader which documents are more relevant than others, even if the reader would not be able to quantify a "relevance score" for each document.

By training a model to identify a more relevant document, a ranking can be applied to the data.

An example of this might be identifying documents in a companies interwiki that are relevant for new employees, by training the model to rank whether one document is more relevant than an other, ultimately an ordered list of documents most relevant for new employees could be created.

3 Implementation

To implement the RankNet approach, a neural network will be trained to identify a score.

Then a sigmoid function to identify a similarity score.

First apply to an obvious data set.

Then use a different optimiser Then allow it to use more data then a different data set Then a non-obvious data set, like wine Finally Documents.

- 3.1 Blobs
- 3.2 Moons
- 3.3 Optimisers
- 3.4 Batches
- **3.5** Wine
- 3.6 Rank Wiki Articles
- 4 Conclusion

5 Further Research

- Apply this to documents to get a sorted list.
- The "Quicksort" algorithm likely needs a random pivot to be efficient [9]

6 Text and References

Fractals are complex shapes that often occur from natural processes, in this report we hope to investigate the emergence of patterns and complex structures from natural phenomena. We begin with an investigation into fractals and the concept of dimension and then discuss links between fractal patterns and natural processes.

This is a Reference [10] and another [8] and yet another [1].

7 Fractals

Images are shown in figure 1.

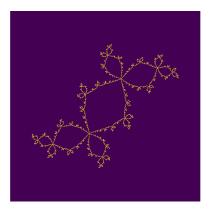


Figure 1: This is a test image showing the outline of a Julia set

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