

#### DEPARTMENT OF ENGINEERING MATHEMATICS

## Improving Text Classifier Performance through Human-in-the-Loop: Enhancing Learning from Explanations

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A dissertation submitted	to the University of Bristol in accordance with the requirements of the degree of Master of Science in the Faculty of Engineering.
	Sunday 13 <sup>th</sup> August, 2023

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## **Declaration**

This dissertation is submitted to the University of Bristol in accordance with the requirements of the degree of MSc in the Faculty of Engineering. It has not been submitted for any other degree or diploma of any examining body. Except where specifically acknowledged, it is all the work of the Author.

Xinyang Song, Sunday 13<sup>th</sup> August, 2023



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## Abstract

Text classification utilises Natural Language Processing (NLP) techniques to analyse pre-trained texts and assign them appropriate labels. Especially in crisis situations like floods, earthquakes, etc., text classifiers are crucial in identifying key information and effectively forwarding it to relevant agencies. However, the efficacy of text classification largely depends on abundant rich training data, which may be difficult to obtain in many scenarios [3]. Relying on a vast amount of non-representative annotated data can lead to delays in project commencement and may impact the model's accuracy. Specifically, in emergencies, identifying action-related information (such as casualties or missing persons) becomes particularly challenging.

To address these issues, this paper introduces a Human-in-the-Loop (HITL) system integrated into the classifier's training process. Coupled with the representational engineering technique of natural language explanations—ExpBERT, a BERT fine-tuned on the MultiNLI natural language inference dataset is utilised to learn from explanations [4]. This optimised embedding representation is used as input to the neural network classifier, further enhancing performance. The core task of the paper revolves around improving the performance of the text classifier based on ExpBERT by combining a limited but representative set of samples with labels and explanations provided by either humans or OpenAI [1].

This paper delves deeply into the application of active learning in text classification tasks based on ExpBERT. During the iterative process, acquisition functions based on uncertainty and diversity are used to select representative unlabeled instances for annotators to process. The original small sample and newly annotated data are used for model retraining. The results indicate that this method can achieve performance similar to models trained on vast datasets within a few iterations. To substantiate this, the model's performance using different active learning strategies is compared on the CrisisNLP dataset, and the effects of an active learning system combined with Monte Carlo Dropout (MCD) are also evaluated.

The main conclusions of this paper are:

- The Human-in-the-loop framework combined with a text classifier enables the model to achieve, or even surpass, the performance of models trained with the full dataset, using only a small amount of annotated data and undergoing a limited number of iterations.
- Compared to merely annotating labels, low-quality explanations, or a limited number of explanations, annotating a certain amount of high-quality explanations can significantly boost the model's performance.
- In terms of acquisition functions, semantic diversity-based sampling and Bayesian Active Learning by Disagreement (BALD) strategies combined with Monte Carlo Dropout (MCD) can allow the model to achieve higher average performance post-training.



# Supporting Technologies

- $\bullet$  I used Python as the development language.
- I used the *Transformers* library to introduce pre-trained models and *Pytorch* for implementing neural networks.
- $\bullet$  I used OpenAI models as one of the annotation methods.
- I used LATEX to format my thesis, via the online service Overleaf.



# Notation and Acronyms

NLP : Natural Language Processing

 ${\bf BERT} \qquad : \quad {\bf Bidirectional \ Encoder \ Representations \ from \ Transformers}$ 

ExpBERT : Representation Engineering with Natural Language Explanations

NLI : Natural Language Inference

 $\begin{array}{lll} \text{HITL} & : & \text{Human-In-The-Loop} \\ \text{AL} & : & \text{Active Learning} \end{array}$ 

BALD : Bayesian Active Learning by Disagreement

MCD : Monte Carlo Dropout

GPT : Generative Pre-trained Transformer

NN : Neural networks



# Acknowledgements

Xinyang Song would like to extend her profound gratitude to her supervisor, Dr. Edwin Simpson, for sharing literature and implementation examples related to ExpBERT and active learning, offering countless invaluable assistance. His professional guidance and enthusiastic support helped me overcome numerous challenges, complete my dissertation, and achieve meaningful insights.

Additionally, I'd like to thank the University of Bristol and my peers for their resources and emotional support during this time. My gratitude goes out to all of you.



### Introduction

This chapter begins by introducing the research background of the text classifier based on ExpBERT integrated with Human-In-The-Loop (Section 1.1). Based on this background, the motivation for the experiment was formed in Section 1.2, emphasising the significance of this research. Furthermore, a general overview of the work will be provided, contrasting with the limitations of traditional methods. Finally, the primary objectives and challenges are briefly summarised (Section 1.3).

#### 1.1 Background

Emergency response systems for social platforms mainly focus on developing better text classification algorithms to learn from limited data. However, obtaining valuable annotated datasets can prove to be challenging [5]. The text classifier based on ExpBERT receives classification labels and corresponding explanations. These explanations detail which keywords led to a particular classification. ExpBERT incorporates this information into the model's training, allowing it to learn from deeper semantic insights to enhance its generalisation capabilities. Therefore, annotating with representative data and understanding the key phrases is vital for guiding the model towards accurate classification.

However, the data throughput of typical social platforms does not permit the annotation of tens of thousands of entries. Consequently, various forms of active learning that can sample high-information-value data at low costs have found extensive application in classification projects [6, 10]. Active learning emerges as an effective solution to these challenges, selecting a limited number of highly informative unannotated samples for human experts in the loop to annotate [9, 2]. The newly data, once effectively annotated, dictates the performance of the model in the next iteration. Thus, employing different acquisition functions to query the most information-rich new instances is perhaps the most popular approach in active learning. Acquisition functions have naturally become the focal point of research in the domain of active learning.

Additionally, neural network-based text classifiers are often ill-suited for early uncertainty sampling [8, 7]. This is because the weight parameters of the neural network are fixed values, causing the model to be overly confident in its predictions, both correct and incorrect. However, by employing dropout, it's possible to introduce a degree of uncertainty into the model. During training, the dropout layer randomly "turns off" a subset of input units. This randomness ensures that network weights are no longer fixed values, thus simulating weight uncertainty.

#### 1.2 Motivation

In light of the aforementioned background, this section emphasises the research motivation from three aspects: accuracy, representativeness, and robustness.

#### 1.2.1 Accuracy

A unified feature of social networks, exemplified by Twitter, is the vast amount of data combined with internet-specific language syntax. The ultimate goal is to precisely classify urgent messages when capturing emergent needs and to dispatch relevant departments to address those needs. However, traditional text classification systems that achieve performance improvements are primarily done through extensive supervised learning, which is costly and slow to respond. Texts with varying amounts of information are randomly distributed, and the difference in information content can significantly impact model accuracy. Consequently, the results of capturing and identifying crucial emergency information are often subpar. Therefore, employing active learning models to utilise a small amount of data and enhance accuracy is urgently needed.

#### 1.2.2 Representativeness

The ability to query rich feature representations from a large pool of unannotated data can mitigate much of the bias in multi-class active learning problems for emergency scenarios. By actively selecting samples for annotation, the model can choose the most valuable and representative samples in each iteration to enhance its performance. Therefore, selecting an effective acquisition function to seek out representative data has the most profound impact on performance and offers the most significant room for improvement.

#### 1.2.3 Robustness

Original models often exhibit an overconfidence bias when handling text classification tasks. When encountering unfamiliar content, they make incorrect and unreliable predictions. In many practical applications, such as disaster response and medical diagnostics, erroneous predictions can have severe consequences. Thus, introducing a certain level of uncertainty to the model to curb its overconfidence has become an essential research topic. On the other hand, we can enhance the robustness of the model without introducing excessive computational overhead. This holds significant value for neural networks in scenarios that demand highly accurate and reliable predictions.

#### 1.3 Experimental Methodology

Considering the background and motivations, this paper eventually designs a pool-based Human-In-The-Loop active learning system for text classification. Unlike traditional text classification models, we train with a small amount of annotated data and then use the trained model in combination with various acquisition functions to select instances. Subsequently, preset annotations or human or OpenAI annotators observe and analyse the key features of the extracted instances in the loop, providing corresponding labels and adding a certain amount of explanations. These newly annotated instances are incorporated into the original data, and the above cycle is repeated until the performance reaches that of the full data (with the training set ratio close to 1, annotated with labels and default explanations).

The implementation will utilise uncertainty sampling based on Least Confidence, sentiment diversity sampling, and random sampling as a baseline to examine the performance under various active learning strategies. Detailed explanations of the acquisition functions will be provided in Chapter 2. Furthermore, in every iteration, the concatenated explanatory texts will be processed using a pre-trained model based on ExpBERT to generate deep semantic representations. This embedded vector will then be input into the neural network text classifier.

Lastly, structural modifications will be made to traditional neural network(NN) models. Using the dropout mechanism, the model becomes more robust to minor variations in input, thereby enhancing its generalisation performance. Additionally, dropout can address the model's overconfidence issue by introducing noise and randomness, ensuring that model predictions aren't overly deterministic. While dropout doesn't directly quantify uncertainty, the introduced randomness and noise can bolster the model's resilience and to some extent alleviate its overconfidence. Given the high computational cost of entropy calculations in the original Bayesian Active Learning by Disagreement (BALD) algorithm, we've refined

the BALD algorithm. Utilising Monte Carlo Dropout (MCD), we employ Least Confidence to select samples where the model's predictive probability is most uniform (i.e., without a particularly high predictive probability for any category).

#### 1.3.1 Objectives

Based on the above experimental descriptions, the objectives of this paper can be summarised as follows:

- Develop a pool-based Human-In-The-Loop framework applied to the ExpBERT-based text classifier and investigate its effectiveness.
- Construct an annotator simulation process and design stopping criteria (e.g., terminate when reaching the performance of the full data model).
- Establish a baseline (performance of the full-data model without active learning or random sampling) and study whether active learning can enhance the performance of the text classification model. Compare the effects of different acquisition functions.
- Explore the impact of the quality and quantity of explanations on this system. For instance, providing explanations with noise or increasing the number of explanations added in each iteration.

#### 1.3.2 Challenges

The main challenges of this project lie in the significant differences that may exist between various active learning strategies and how to improve the original initialisation framework of the multi-classifier based on ExpBERT and the annotation process to reduce computational costs.

- It's worth noting that there are multiple acquisition functions in active learning text classification systems. The application of different algorithms in active learning strategies significantly affects the performance of the active learning framework. The choice of strategy needs to consider both time complexity and representativeness.
- Secondly, neural networks with dropout mechanisms might increase computational costs during training algorithm execution and when using MCD for active learning. Thus, how to reduce computational costs in each iteration to improve response efficiency poses a challenge.
- Finally, regarding the framework design, determining the appropriate sampling ratio, the number of iterations, and the number of explanations provided in each iteration requires further study to achieve satisfactory performance.

CHAPTER	1	INTRODUCTION
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## Background

A compulsory chapter, roughly 20% of the total page-count

This chapter is intended to describe the technical basis on which execution of the project depends. The goal is to provide a detailed explanation of the specific problem at hand and existing work that is relevant such as an existing algorithm that you use, alternative solutions proposed, supporting technologies, and relevant literature. The literature you include should cover appropriate peer-reviewed academic publications and books, and maybe also news and current-affairs articles published in reputable sources such as *The Economist* magazine or *The Financial Times* newspaper.

Your thesis should include complete set of references/bibliography: everything that you cite should be in there, in full. This means including the publisher's name for anything that is a book; the editors and title of books that a paper appears in as one item in a larger collection (e.g. proceedings volumes and/or thematic edited collections) and the relevant page-numbers; and so on. Put most simply, we require and expect your references to look like the references in a published peer-reviewed academic paper, so you can work out whether you still have work to do by comparing your references to the reference-list on the publications that you're working from.

Note there is a subtle difference from this and a full-blown literature *survey*. The latter might try to capture and organise (e.g., categorise somehow) *all* related work, potentially offering meta-analysis, whereas here the goal is simply to ensure that your thesis is self-contained. Put another way, after reading this chapter an intelligent non-expert reader with no prior knowledge of your project should have obtained enough background to understand what *you* have done (by reading subsequent sections), and then accurately assess your work. You might view an additional goal as giving the reader confidence that you are able to absorb, understand and clearly communicate highly technical material.

Just as there is no single ideal structure for an MSc thesis, there is no one correct name for this chapter. You could just call it *Background* if you wish, or *Technical Background*. Or if you feel you have a lot to say you could split this chapter into two: you might have your first three chapters being:

- 1. Introduction
- 2. Context
- 3. Related Work

But if you prefer to use a more concise structure, you could instead use:

- 1. Introduction: Contextual Background
- 2. Technical Background

The choice is yours.

The key thing is that by the end of these first two or three chapters you have told the reader everything they need to know so that they can understand the rest of your thesis. This is particularly important because at least one of the people who actually examines your thesis will be a UoB academic, one of our lecturers or professors, who you can assume knows almost nothing about the specifics of what work you have done, and who is given a copy of your thesis to mark. So what you write has to adequately explain things to that examiner. You can safely assume that whoever examines your thesis is numerate,

intelligent, understands programming and data analytics etc, but you cannot safely assume that their specific individual expertise is a perfect match to the topic of your thesis (it's in that sense that the examiner is a *non-expert*). So, the person you write for, your *audience*, should be that undefined group of academics who might possibly examine your thesis: if you write for that audience and do a good job, your thesis should be understandable by a wide range of people, including potential employers and colleagues in the world of work.

### Execution

A topic-specific chapter, roughly 30% of the total page-count

This chapter is intended to describe what you did: the goal is to explain the main activity or activities, of any type, which constituted your work during the project. The content is highly topic-specific. For some projects it will make sense to split the content into two main sections, or maybe even into two separate chapters: one will discuss the design of something, including any rationale or decisions made, and the other will discuss how this design was realised via some form of implementation. You could instead give this chapter the title "Design and Implementation"; or you might split this content into two chapters, one titled "Design" and the other "Implementation".

Note that it is common to include evidence of "best practice" project management (e.g., use of version control, choice of programming language and so on). Rather than simply a rote list, make sure any such content is useful and/or informative in some way: for example, if there was a decision to be made then explain the trade-offs and implications involved.

### 3.1 Example Section

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**Algorithm 3.1:** This is an example algorithm.

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for( i = 0; i < n; i++ ) {
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Listing 3.1: This is an example listing.

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**Example paragraph.** This is an example paragraph; note the trailing full-stop in the title, which is intended to ensure it does not run into the text.

## Critical Evaluation

A topic-specific chapter, roughly 30% of the total page-count

This chapter is intended to evaluate what you did. The content is highly topic-specific, but for many projects will have flavours of the following:

- 1. functional testing, including analysis and explanation of failure cases,
- 2. behavioural testing, often including analysis of any results that draw some form of conclusion wrt. the aims and objectives, and
- 3. evaluation of options and decisions within the project, and/or a comparison with alternatives.

This chapter often acts to differentiate project quality: even if the work completed is of a high technical quality, critical yet objective evaluation and comparison of the outcomes is crucial. In essence, the reader wants to learn something, so the worst examples amount to simple statements of fact (e.g., "graph X shows the result is Y"); the best examples are analytical and exploratory (e.g., "graph X shows the result is Y, which means Z; this contradicts [1], which may be because I use a different assumption"). As such, both positive and negative outcomes are valid if presented in a suitable manner.

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## Conclusion

#### A compulsory chapter, roughly 10% of the total page-count

The concluding chapter(s) of a dissertation are often underutilized because they're too often left too close to the deadline: it is important to allocate enough time and attention to closing off the story, the narrative, of your thesis.

Again, there is no single correct way of closing a thesis.

One good way of doing this is to have a single chapter consisting of three parts:

- 1. (Re)summarise the main contributions and achievements, in essence summing up the content.
- 2. Clearly state the current project status (e.g., "X is working, Y is not") and evaluate what has been achieved with respect to the initial aims and objectives (e.g., "I completed aim X outlined previously, the evidence for this is within Chapter Y"). There is no problem including aims which were not completed, but it is important to evaluate and/or justify why this is the case.
- 3. Outline any open problems or future plans. Rather than treat this only as an exercise in what you could have done given more time, try to focus on any unexplored options or interesting outcomes (e.g., "my experiment for X gave counter-intuitive results, this could be because Y and would form an interesting area for further study" or "users found feature Z of my software difficult to use, which is obvious in hindsight but not during at design stage; to resolve this, I could clearly apply the technique of Bloggs et al..

Alternatively, you might want to divide this content into two chapters: a penultimate chapter with a title such as "Further Work" and then a final chapter "Conclusions". Again, there is no hard and fast rule, we trust you to make the right decision.

And this, the final paragraph of this thesis template, is just a bunch of citations, added to show how to generate a BibTeX bibliography. Sources that have been randomly chosen to be cited here include:

## Bibliography

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## Appendix A

# An Example Appendix

Content which is not central to, but may enhance the dissertation can be included in one or more appendices; examples include, but are not limited to

- lengthy mathematical proofs, numerical or graphical results which are summarised in the main body,
- sample or example calculations, and
- results of user studies or questionnaires.

Note that in line with most research conferences, the examiners are not obliged to read such appendices.