

WUST at the NTCIR-14 FinNum Task

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Introduction

In NTCIR-14, the FinNum task is dedicated to identifying the

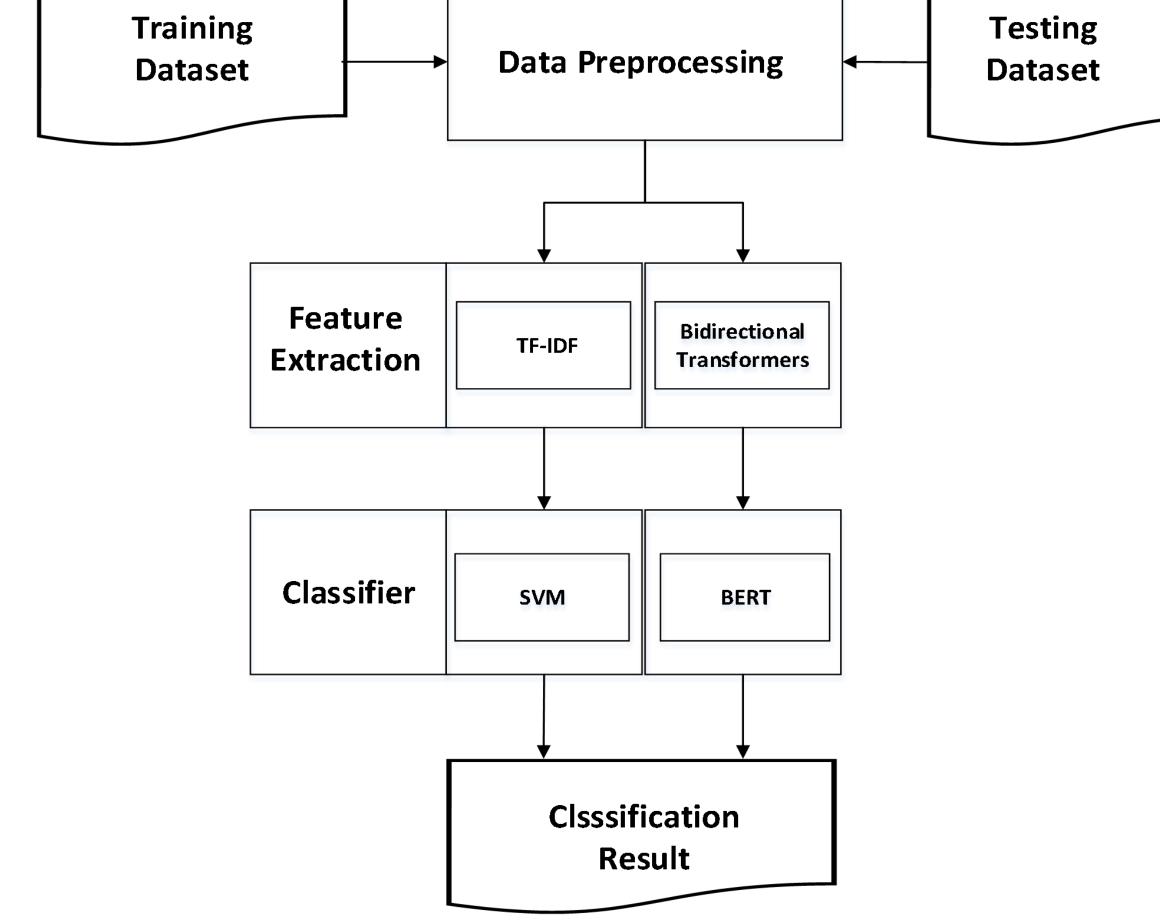
3. SVM Classifier

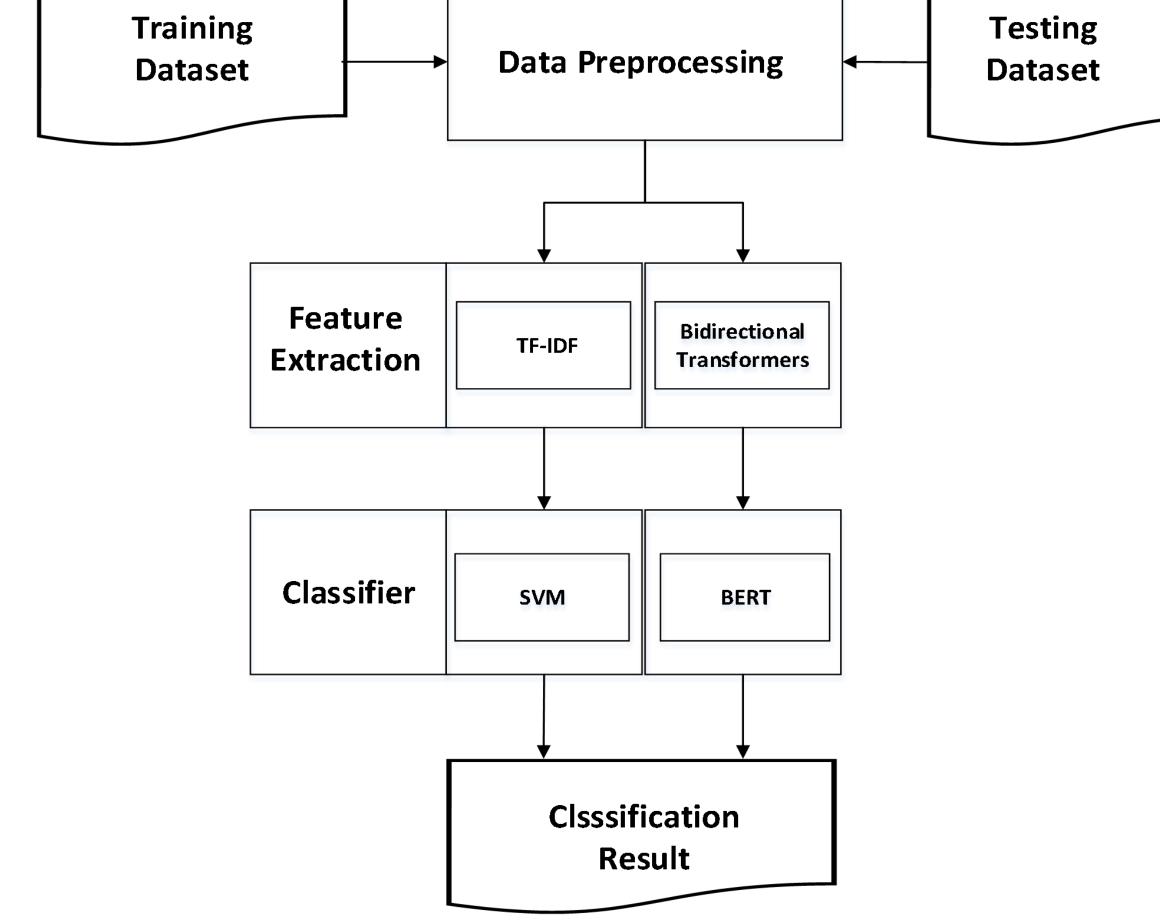
The SVM model has the characteristic of high precision and

- category of a numeral from the financial social media data, i.e. tweet, for fine-grained numeral understanding.
- Our team carry out the FinNum task as the text classification problem. The Subtask1 is one seven-classification task, and the subtask2 is one seventeen-classification task.
- In our system, we construct the classification model based on Support Vector Machines (SVM) to identify the categories of numerals. In additional experiments, we adopt the Bidirectional Encoder Representations from Transformers (BERT) model to act as a multi-classifier.
- The experimental results show that our proposed both SVM and BERT models are effective.

System Architecture

•Our system consists of three main modules, i.e. data preprocessing, feature extraction, and classifier.





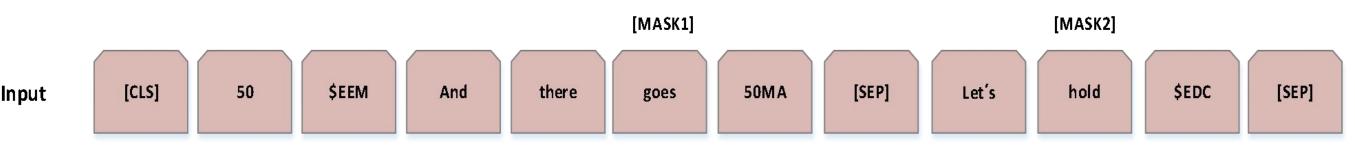
multidimensional data processing, and strong robustness, so the SVM model is favored in the study of classification problems. The SVM is usually used for binary classification problems, but the FinNum task is a multi-classification problem, so a multiclassification SVM needs to be constructed.

The specific formula is shown as follows.

$$k(x, x_i = exp(-\gamma ||x_i - x_j||^2)), \gamma > 0$$

4. BERT Classifier

The BERT uses a simple approach, in which a special classification embedding ([CLS]) inserted as the first token and a special token ([SEP]) added as the final token. It masks out 15% of the words in the input, runs the entire sequence through a deep bidirectional Transformer encoder, and then predict only the masked words. the Sentence Embedding as shown in Fig. 1.



1. Data preprocessing

 \bullet In the data preprocessing, due to the uneven data distribution, our team tries to expand the training data, ensuring that one target numeral corresponds to one tweet and one category, divide tweets with the target numeral, and combine the segmented data into the training set. As a result, the divided data are tripled.

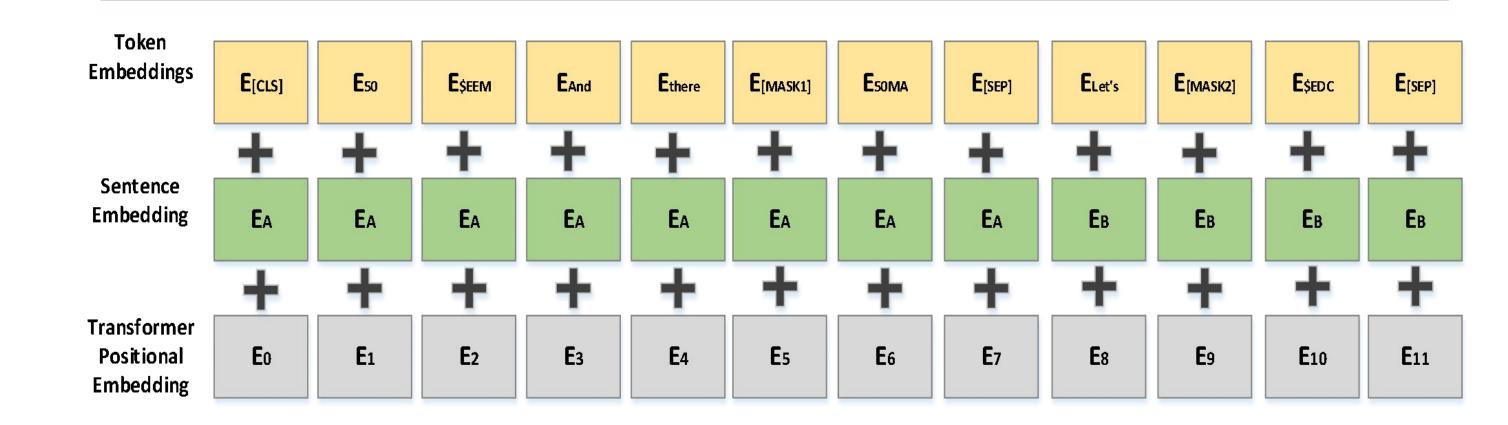


Fig 1. Sentence Embedding

Experiments

• We submitted one system result to NTCIR-14 for the FinNum task. The official evaluation results are listed in Table 1

Task Name	Micro	Macro
Subtask1	0.7402	0.6371
Subtask2	0.6088	0.5293

Table 1. Performances of the SVM model

In additional experiments, we selected the BERT model as a

2. Feature extraction

Before classifying with SVM, we extract textual features by Term Frequency-Inverse Documentation Frequency (TF-IDF), and TF-IDF is a statistical method used to evaluate the importance of a word to a document set or one of the documents in the corpus. The TF-IDF algorithm is simple in principle, fast in the calculation, and has strong universality for extracting keywords for FinNum task.

In additional experiments, we adopt the BERT model, which trains a universal language understanding model on a large-scale corpus, a deep bidirectional Transformer encoder is used to retain contextual features and positional information when extracting textual features.

classifier. The experimental results are shown in Table 2.

Task ·Name	Micro	Macro
Subtask1	0.9450	0.8862
Subtask2	0.8725	0.8307

Table 2. Performances of the BERT model

Conclusions

• we separately employ SVM and BERT models for classification, both of which have got good performances. Moreover, we have mainly taken statistical features into consideration, and we will extract and select more rules and semantic features for our system to improve the system accuracy.