

Teamwork Dimensions Classification Using BERT

Junyoung Lee^a and Elizabeth Koh^b

^aNanyang Technological University, Singapore

^bNational Institute of Education, Singapore

1 Introduction

To provide a formative assessment of student teamwork, an automated natural language processing approach was developed to identify teamwork dimensions of students' online team chat, using traditional machine learning methods [1, 2]. The teamwork dimensions are focused on the following areas in Figure 1.

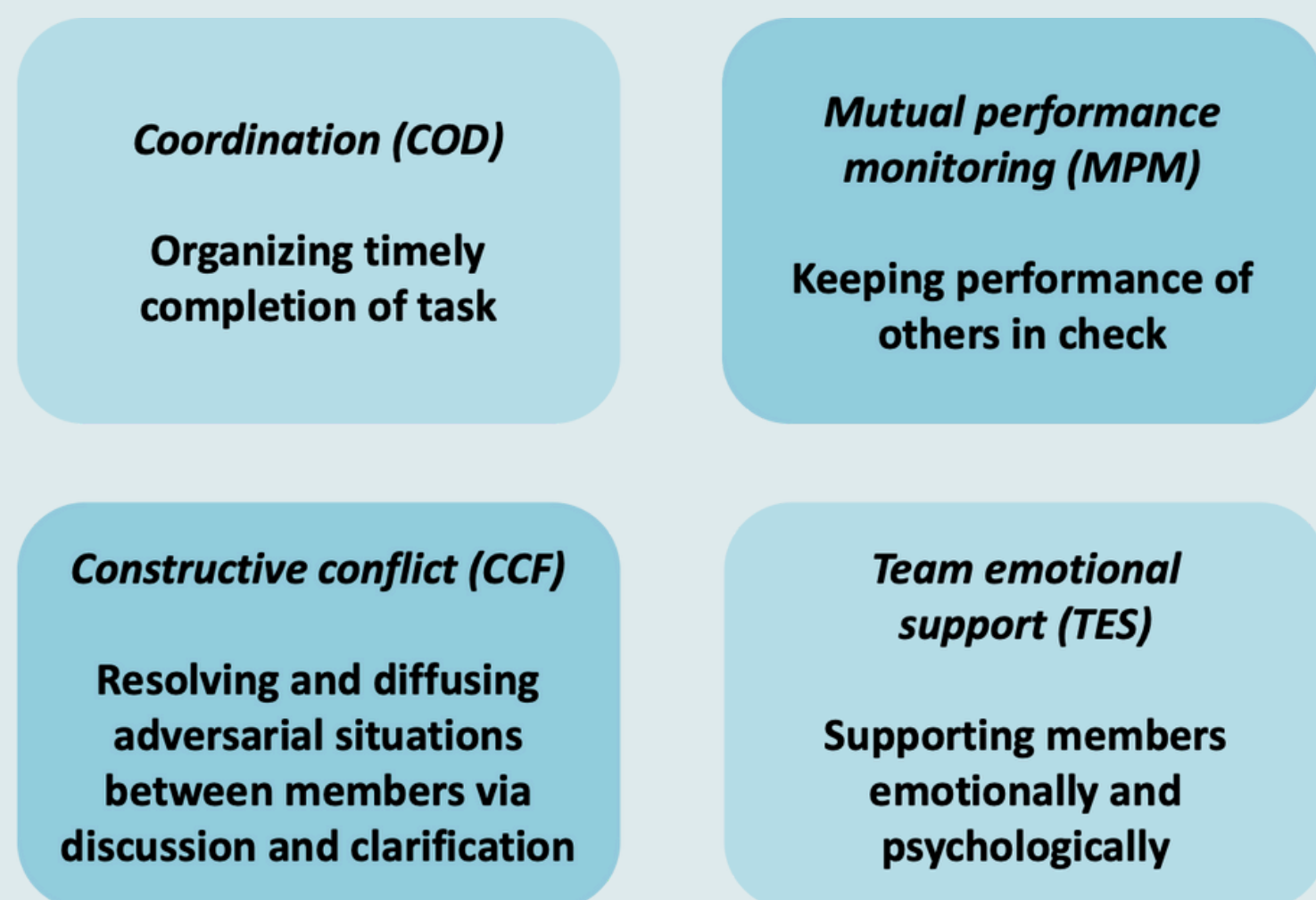


Figure 1. Framework for teamwork competency, derived from [3]

2 Current Challenges

The existing backend model and current training approach face the following challenges:

1. Resource-heavy data labelling
2. Generalizability of machine learning model
3. Improving classification performance

3 Objectives

To improve the existing backend model - random forest (RF) classifier with TF-IDF vectorizer - for teamwork classifications using deep transfer learning techniques, namely Bidirectional Encoder Representations from Transformers (BERT), in terms of classification performance and generalizability.

The approach considers the recent success of pre-trained language models in natural language processing tasks.

4 Dataset and Pre-processing

Chatlog of 19762 messages gathered from 76 teams of a total of 272 14-year-old students in a Singapore secondary school. Table 1 shows a sample of the human-annotated dataset from a past study for the four teamwork dimensions.

Table 1. Sample of annotated dataset

User	Message	COD	MPM	CCF	TES
Student A	Bob are you okay with it	1	1	0	0
Student B	ideal teacher as like um can work well with students and listen to students ideas	0	0	1	0
Student C	yes	0	0	0	1
Student C	So, caring, attentive to our needs and humourous?	0	0	1	0
Student A	humour yes	0	0	0	1
Student B	ok	0	0	0	1

Text pre-processing was carried out, with focus on textese and local terms replacement. Another approach previously taken to improve classification performance was feature engineering on the text data [2], hand-crafted from language cues related to the teamwork dimensions.

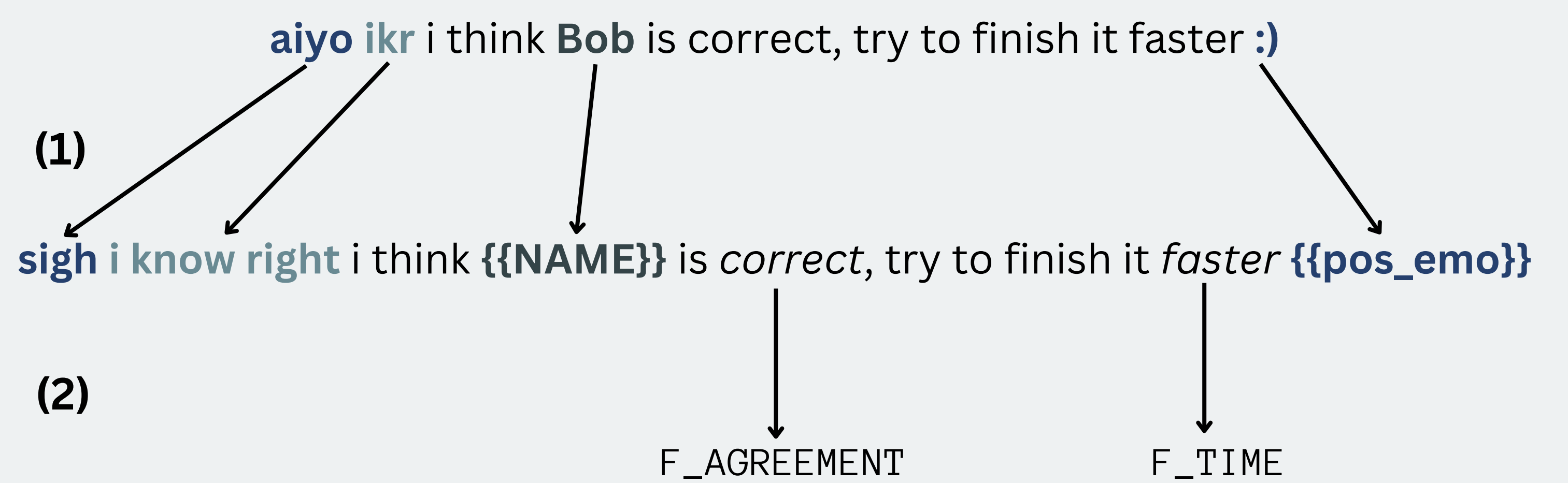


Figure 2. Example of (1) text pre-processing and (2) feature engineering

5 Results

A BERT-based classifier was developed, and its classification performance was compared on both the feature engineered and non-feature engineered versions of the corpus, as seen in Table 2.

Inference on Unseen Data

Chatlog of 129 messages gathered from 2 groups of 15 students aged 19 to 42, manually annotated, pre-processed using the steps in Table 2 with no additional features. The trained BERT-based classifier achieved a Cohen's Kappa score of **0.640** compared to 0.467 achieved by the RF classifier, showing a higher inter-rater reliability.

Table 2. Classification performance comparison

	Without feature engineering		With feature engineering	
	RF	BERT	RF	BERT
Precision	0.824	0.810	0.833	0.801
Recall	0.406	0.702	0.594	0.747
F1 Score	0.527	0.734	0.684	0.757
Hamming distance	0.124	0.076	0.090	0.070

6 Conclusion

The project demonstrates potential for **scalability** of such models **without need for extensive feature engineering**, as well as **generalizability** of the model with different demographics. This ultimately contributes towards an enhanced formative learning analytics tool for teamwork assessment, allowing students and teachers to become more aware of the teamwork competencies in the discussion process and receive more holistic feedback.

Future works include further in-depth analysis of task context, as well as more complex deep learning architecture or advanced language models.

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