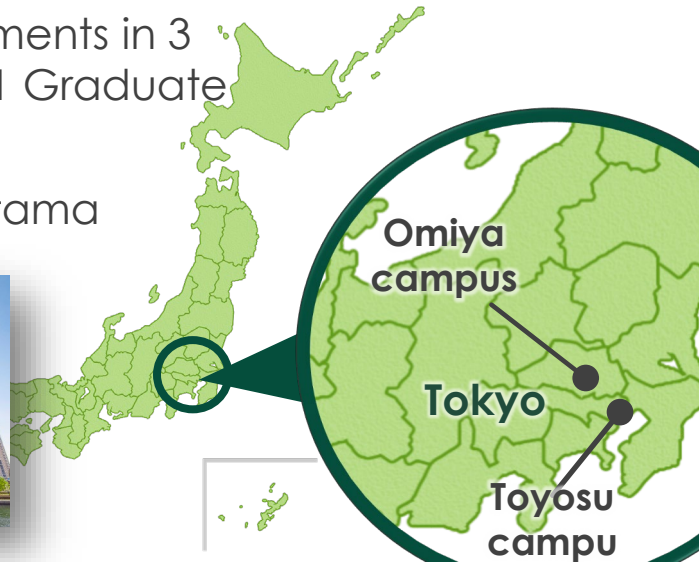


DEEP LEARNING BASED ASSESSMENT IN HIGHER EDUCATION: CASE STUDIES

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Shibaura Institute of Technology (SIT)

- ▶ Private school for engineering education and research
- ▶ 7 Programs and 7 Departments in 3 Colleges and 1 School & 1 Graduate School
- ▶ Located in Tokyo and Saitama



Challenges of assessment based on deep learning

- ▶ Assessment on students' essays
 - ▶ Content evaluation (essay scoring)
 - ▶ Author identification
- ▶ Program assessment for active learning programs (PBL etc.)
 - ▶ Understanding students' discussion
etc.



Hereafter, let me introduce our group's researches related to assessment

A Method for Distinguishing Model Generated Text and Human Written Text

Hinari Shimada, Masaomi Kimura

Data Science/Engineering Laboratory

Shibaura Institute of Technology

Presented at ICIT2024, Ho Chi Minh city, Vietnam

Procedure of Proposed Method

1. Input each i -th text x_i from dataset X into n GPT-based LLMs, and compute n log likelihood $\log p_1(x_i), \dots, \log p_n(x_i)$
2. Input X into m BERT-based LLMs to obtain 768-dimensional vectors, followed by dimensionality reduction to 2 dimensions using t-SNE
3. Input 1 and 2 into a feedforward neural network (FFNN) for classification

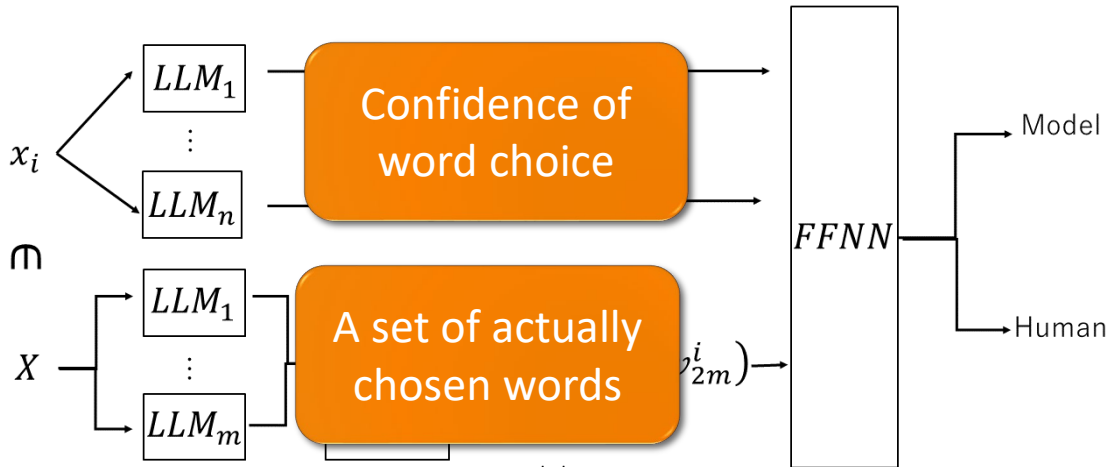


Fig.2 : Model Overview

Settings and Results

● Selection criteria of LLMs

- Log likelihood calculation : Four GPT-based LLMs
 - Based on previous research[5] suggesting that more advanced GPT-based models tend to use sophisticated vocabulary and syntax
- Embedding vector calculation : Four BERT-based LLMs
 - Selecting four standard base version models

● Result

- **Best accuracy** compared to prior research

Table 1 : Result

Model	Accuracy [%]
DetectGPT	70.48
Roberta-HPPT	86.60
CheckGPT	97.95
Proposed Model	98.49



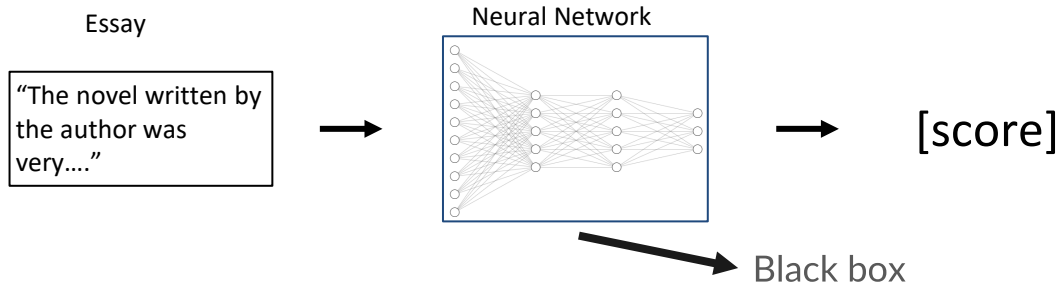
Enhancing Automated Essay Scoring Model's Interpretability Using Triplet Loss with Teacher-Defined Rubric as Anchors

The 6th Joint International Conference on AI, Big Data and Blockchain
Istanbul, Türkiye

Joseph Saido Gabriel and Masaomi Kimura
Shibaura Institute of Technology, Tokyo, Japan

Introduction | What is Automated Essay Scoring?

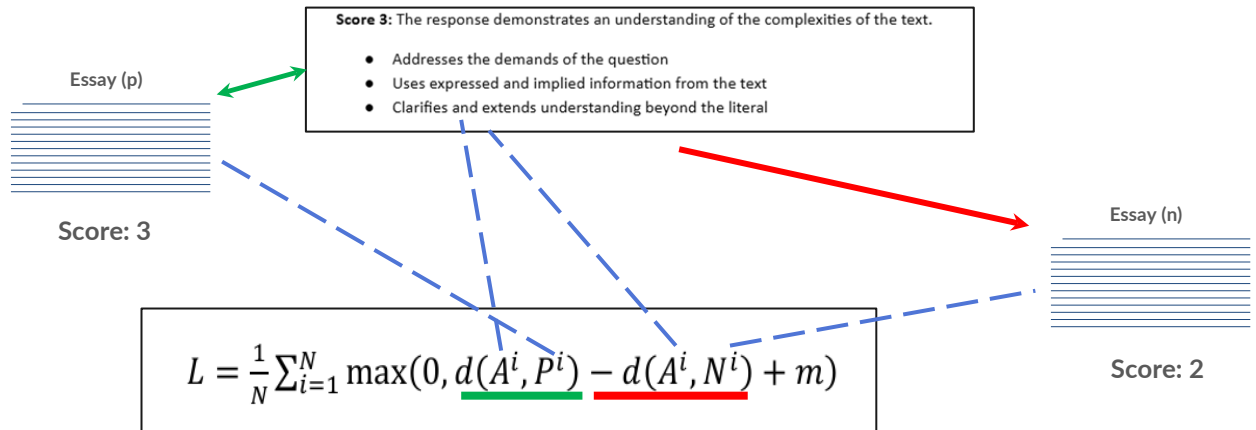
Automated Essay Scoring (AES) is a task that utilizes Natural Language Processing (NLP) techniques to assign scores to essays written by students.



Methodology

- We propose a novel rubric-aligned AES model using triplet loss and dual-anchor triplet loss
 - Triplet loss is a key technique in metric learning, aiming to create an embedding space where similar items are closer and dissimilar ones are farther apart.

Since this essay has score three, the model will learn to structure the **essay's embedding** similarly to the corresponding rubric's embedding.



Experiment Result | QWK Scores

Models	Set 1	Set 2	Set 3	Set 4	Set 5	Set 6	Set 7	Set 8	Avg.	Explainability
R ² BERT [6]	0.817	0.719	0.698	0.845	0.841	0.847	0.839	0.744	0.794	✗
NPCR [11]	0.856	0.750	0.756	0.851	0.847	0.858	0.838	0.779	0.817	✗
GPT-4, zero-shot, w/o rubrics [13]	0.042	0.402	0.281	0.557	0.366	0.502	0.081	0.419	0.331	✓
GPT-4, zero-shot, with rubrics [13]	0.072	0.300	0.366	0.627	0.523	0.345	0.110	0.407	0.344	✓
GPT-4, few-shot, with rubrics [13]	0.280	0.338	0.331	0.784	0.623	0.728	0.257	0.454	0.474	✓
Fine-tuned GPT-3.5 [13]	0.741	0.618	0.704	0.859	0.796	0.848	0.727	0.614	0.738	✓
Holistic rubric model	0.810	0.710	0.735	0.836	0.822	0.836	-	-	0.770	✓
Multiple trait rubric model	-	-	-	-	-	-	0.730	0.677		✓

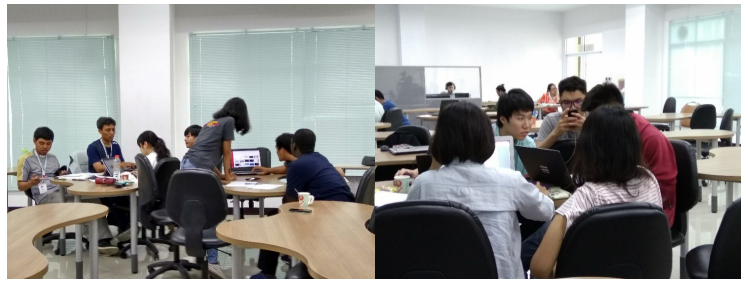
Deep Learning Based Text Mining on Global PBL Description Data

Masaomi Kimura, Hiroyuki Ishizaki, Hitoshi Nakamura, Tomoko Iwata
Shibaura Institute of Technology, Japan

Maria Anityasari, Imam Mukhlash, Mohammad Iqbal¹
Institut Teknologi Sepuluh Nopember, Indonesia

Presented at ICIIT 2025, Hanoi, Vietnam

Global PBL



- Global PBL is a student-centered learning activity program
- These programs often involve international and interdisciplinary collaboration
- However, **assessing these** programs to facilitate continuous improvement **remains challenging**.
 - Necessary to analyze daily reports by participants

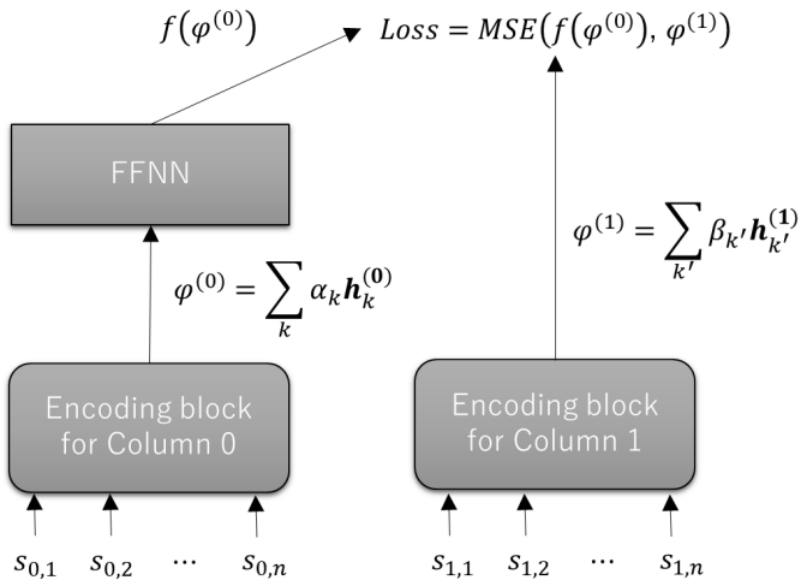
Name: _____		
Day & Date: _____		
Topic: _____		
Curiosity (Fill in before you listen to the presentation of the speakers)	Interesting Facts or Information (Fill in during the presentation of the speakers)	Identified Problems or Needs (Fill in during the presentation of the speakers)
Data or Information to be Further Explored (Fill in during or after the presentation of the speakers – before you discuss with your group mates)		
Points of Discussion with Group Members (Fill in after the presentation of the speakers)	Further Questions for the Feasibility Study (Fill in after the presentation of the speakers)	

Our proposed model

Adjustment of
attention
weights α

based on the relationships
of description

Text encoding
by Sentence
BERT



An analysis example: for gPBL whose topic was environment eng.

Curiosity

0:clean energy

3:biomass

5:products

6:traffic

2:waste

1:space data/
air pollution

4:policy

Interesting facts and info

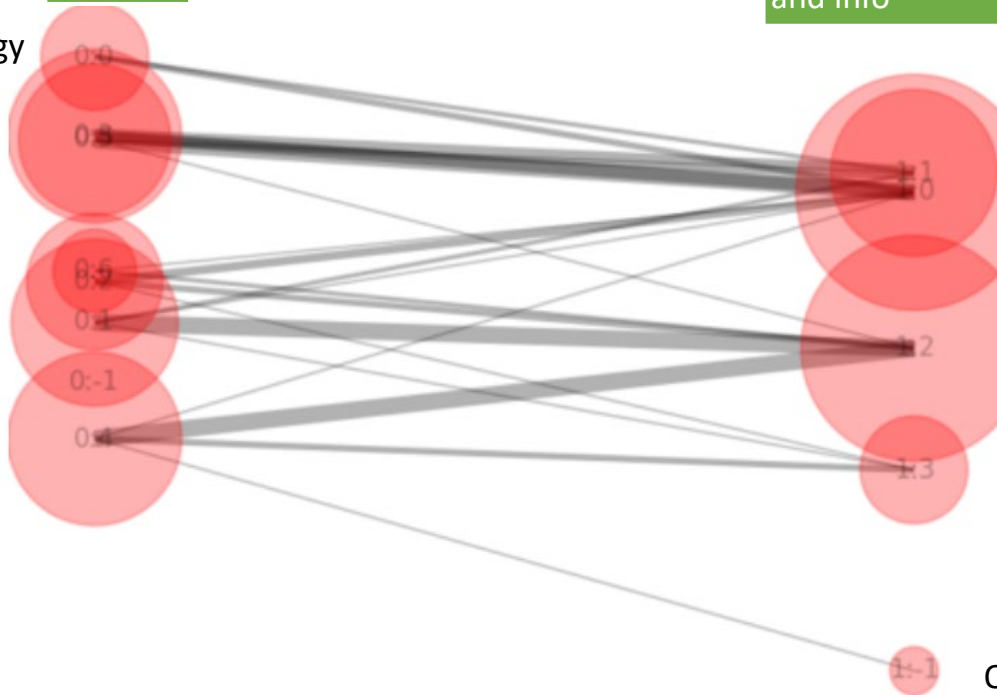
0:zero
emission

1:env. in
Indonesia

2:waste-
electricity
conversion

3:waste
landfill
company

Outlier



Thank you for your attention



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