

# Identifying PII in Educational Datasets

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**Class:** W266 - Natalie (Wednesday)

**Term:** Spring 2024

# Overview

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- **Motivating Question:** What is the best solution for identifying personally identifiable information (PII) in student essays?
  - Student data often requires manual labelling
  - Potential for rich data sets collected from educational data
  - **Do state-of-the-art techniques sufficiently identify student PII?**

# Background

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- Named entity recognition (NER) well-researched field
  - Transformers models standard
    - SpaCy provides open-source library for NER
- Identifying *student* PII rather than PII generally
  - Ex: “In 1865, Abraham Lincoln...” vs. “My friend Natalie...”

# Dataset

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- ~7k student essays written by students enrolled in a massively open online course
  - Responding to one single assignment prompt
- Original PII replaced by surrogate identifiers using partially automated process.
- Target PII types: student names and usernames, B-I-O format

My name is **David Gonzalez,** and my

☐ ☐ ☐ ☒ B-Student Name ☒ I-Student Name ☐ ☐

classmate **Adrianna** (**abk24@gmail.com**) and I participated

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B-Student Name

B-Username

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the March 6 data management workshop at Stanford

O O - Date O O O O - Organization

# Challenges

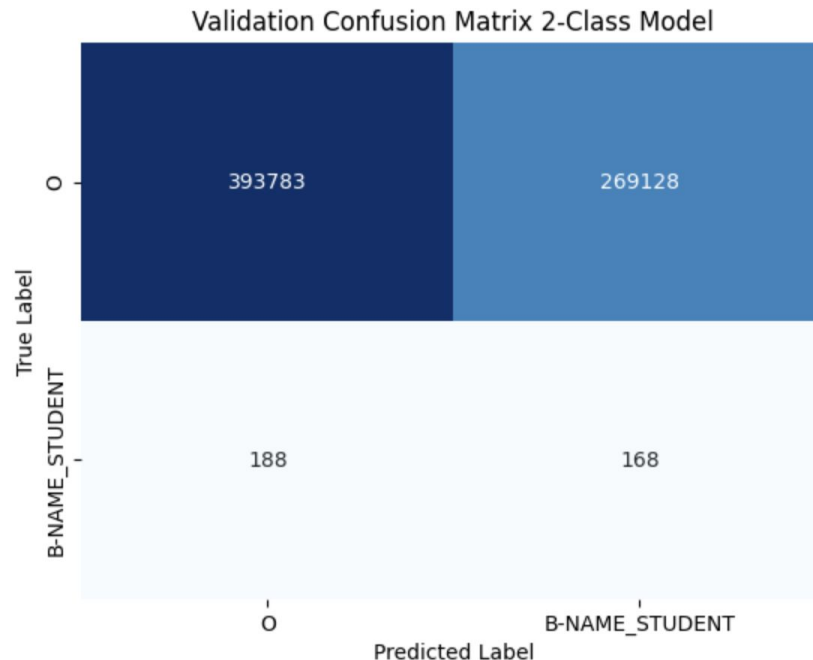
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- Text classification —> Token classification
  - Handling output dimension of:  
(batch size, sequence length, number of classes)
  - Aligning labels with valid text to account for transformers special tokens and padding
- Implementation of weighted loss function to account for class imbalance

# Models and Results

## Weighted F1 Scores of Evaluated Models

- SpaCy: 0.997
- BERT: 0.24
- DeBERTa:
  - Pre-trained model and continue pre-trained model: 0.97
  - Fine-tuned: 0.12
  - Fine-tuned two-class (binary) model: 0.74



# Conclusions

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- NLP systems show great performance detecting student PII
  - SpaCy shows superior performance
  - Pre-trained deBERTa also good performance
- Future directions:
  - Address class imbalances:
    - Weighted loss function
    - Incorporate more PII data
  - Evaluate performance of bidirectional LSTM