



Problem Statement

- Georgia State University and the Learning Agency Lab would like to improve the effective writing skills of students.
- The education system did not put much emphasis in persuasive writing, which may hinder critical thinking development.

Objective:

• Build a model to classify argumentative elements in student writing as "effective", "adequate" or "ineffective".



Data



Data - Discourse Type

Discourse Types

The dataset presented here contains argumentative essays written by U.S students in grades 6-12. These essays were annotated by expert raters for discourse elements commonly found in argumentative writing:

- 1. **Lead** an introduction that begins with a statistic, a quotation, a description, or some other device to grab the reader's attention and point toward the thesis
- 2. **Position** an opinion or conclusion on the main question
- 3. **Claim** a claim that supports the position
- 4. **Counterclaim** a claim that refutes another claim or gives an opposing reason to the position
- 5. **Rebuttal** a claim that refutes a counterclaim
- 6. **Evidence** ideas or examples that support claims, counterclaims, or rebuttals.
- 7. **Concluding Statement** a concluding statement that restates the claims



Data Cleaning

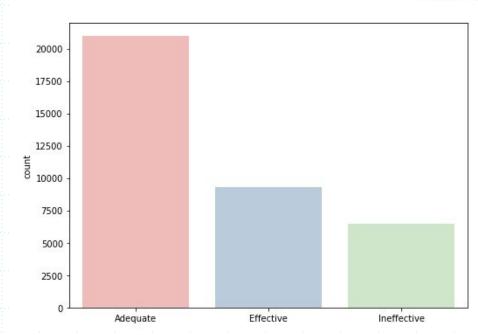
Discourse_id: 451b76cc4b59

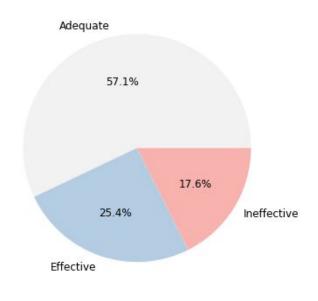
After cleaning

'After all, many students will greduate and keep going with their life without having any problem about jobs and they will be good economically. They also will give their childs a reason why they should go to school an d follow their dreams without letting people to punish their dreams. It also will help their family to support them without caring about what other people says, but everyone should conserve good things about everything. Also this students will have an illustrate part of his life, where they are going from success to success keeping everything on the side but his dreams on the front.

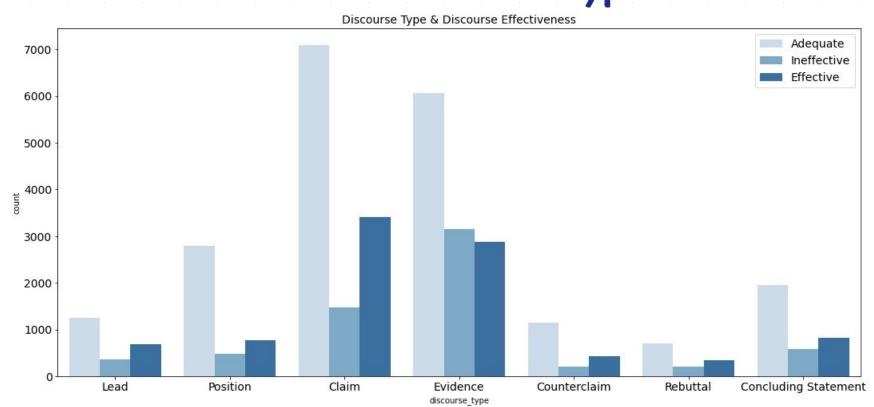
EDA - Target Class

discourse_effectiveness

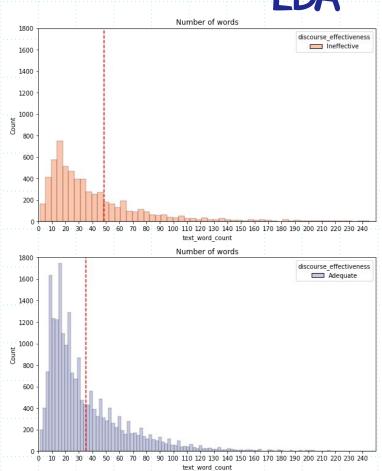


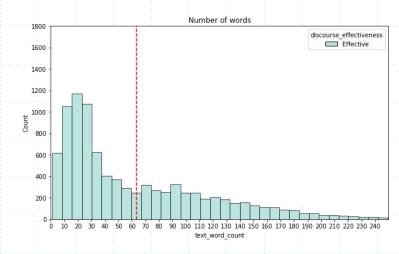


EDA - Discourse Type



EDA - Number of Words





Feature Engineering

1. Discourse <u>Text</u>

2. Discourse <u>Type</u> + Tokenizer Separator + Discourse <u>Text</u>

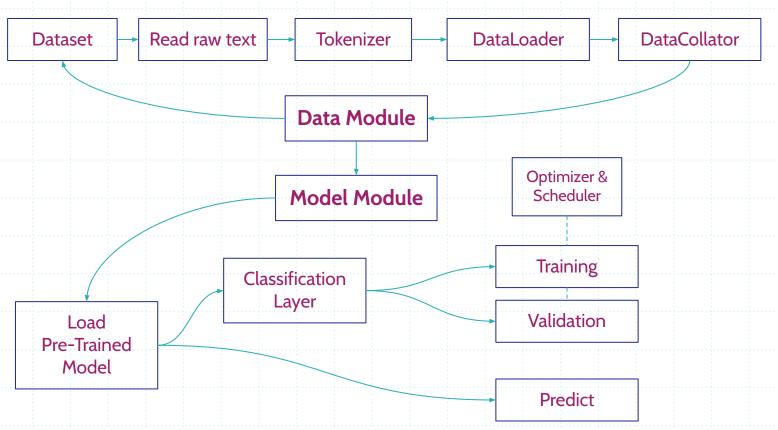
3. Discourse <u>Type</u> + Spacing + Discourse <u>Text</u> + Tokenizer Separator + <u>Essay</u> Text



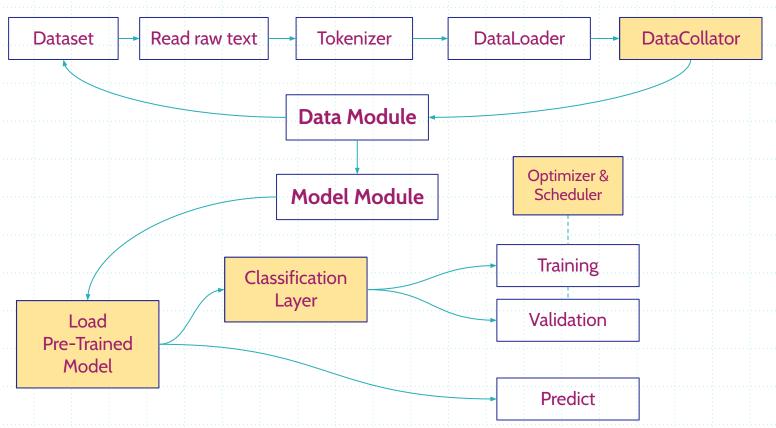
Approach - Fine Tuning PreTrained Models

Fine-tuning with DistilBert Pytorch Ensemble **Bert Pruned** Fine-tuning with Weighted-Average Pytorch **OFA** Fine-tuning with Deberta-v3-Base Pytorch

Modelling Flow

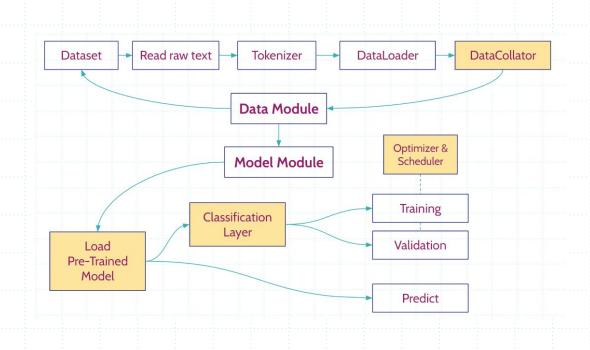


HyperParameter Tuning



HyperParameter Tuning

- 1. Dynamic Padding
- 2. Layer Freezing
- 3. Average Pooling Layer
- 4. Additional Hidden Layer
- 5. Dropout
- 6. Learning Rate
- 7. Warmup

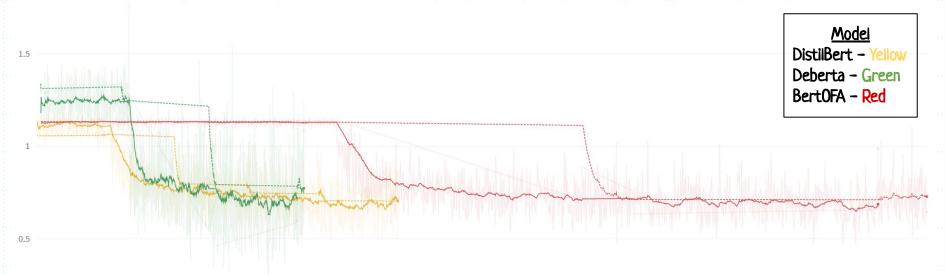


Training Time

Training Loss, Validation Loss

= E3Size16Lr3e-05Warm0.01Weight0.01Freeze2Drop0.1Text0 full Training Loss = E3Size64Lr0.0001Warm0.2Weight1e-06Freeze21Drop0.1 full Training Loss = epoch5,lr6-e5,weight0.001,dropout0.6_embedding&2transformlayerfreeze,fulldata Training Loss = E3Size16Lr3e-05Warm0.01Weight0.01Freeze2Drop0.1Text0 full Validation Loss

== E3Size64Lr0.0001Warm0.2Weight1e-06Freeze21Drop0.1 full Validation Loss == epoch5,1r6-e5,weight0.001,dropout0.6_embedding&2transformlayerfreeze,fulldata Validation Loss



Time (minutes)

20

100



123

Evaluation Metric

Differentiable

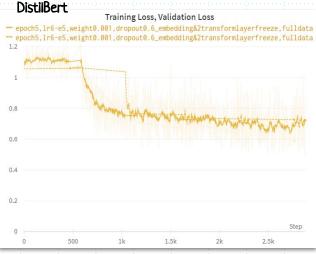
1. Cross Entropy Loss

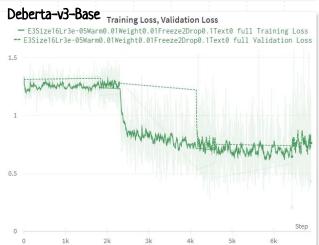
(2.) F1 - Score

Multi-Class Classification

Imbalance dataset



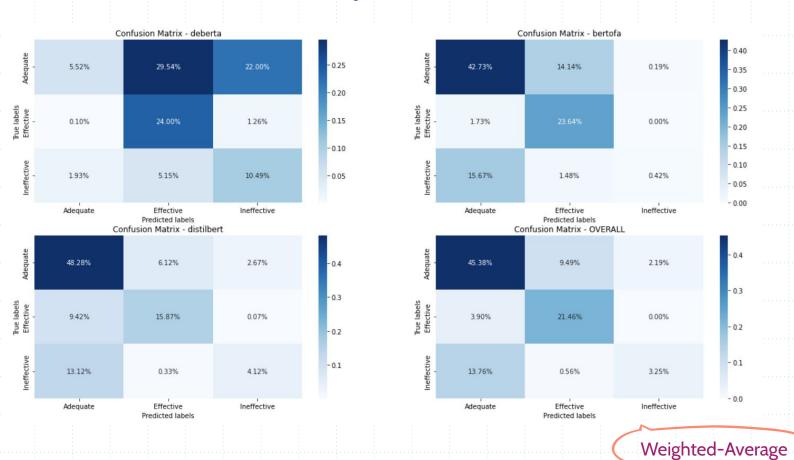








Results





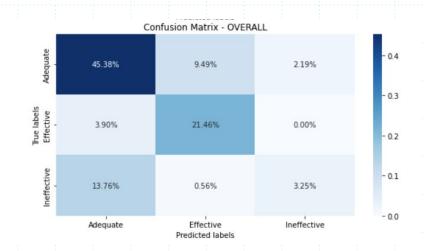
Results

		deberta	distilbert	bertofa	overall
7	label				

Adequate	0.170911	0.754998	0.729256	0.755661
Effective	0.571105	0.665716	0.731481	0.754663
Ineffective	0.408585	0.337416	0.046372	0.282506

- Ensemble technique:
 - a. **Increases** f1 score
 - b. **Reduces** log loss

f1_weighted_s	core	log los	ss model
0.31	4177	1.58375	deberta
0.65	8979	0.71208	88 distilbert
0.60	9831	0.72809	91 bertofa
0.67	2270	0.66811	2 OVERALL







Conclusion

a

Future Work

- Model performs well in classifying "Adequate" and "Effective" labels
- Model unable to classify "Ineffective" label effectively
- Ensemble was able to improve the score but not significantly

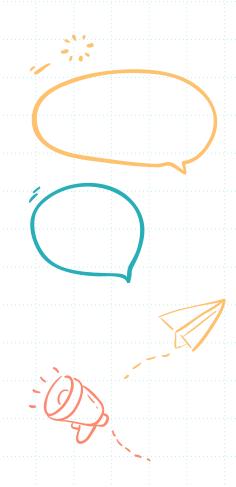
- Review hyperparameter
- Try batch normalization to reduce loss fluctuation during training
- Relook BertOFA model to ensure correct implementation
- Include weights in loss function to counter data imbalance



819 Way Keat Koh

O.710

Your Best Entry!
Your most recent submission scored 0.710, which is the same as your previous score. Keep trying!



Thanks!

Does anyone have any questions?

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