

Return_Zero at Bhashabhrom: Bangla Grammatical Error Detection Leveraging Transformer-based Token Classification

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Errors In Bangla Text

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TABLE I
Types of Spelling Errors

Error Type	Example
Cognitive Error	পরবাস → পরবাশ
Visual Error	দেবতা → দেরতা
Typo Insertion	চুল্লি → চুলল্লি
Typo Deletion	দুর্বার → দুর্বর
Typo Transposition	ঘাটতি → ঘাতটি
Typo (Avro) Substitution	চেয়ার → চেয়াএ
Typo (Bijoy) Substitution	ঘুর্ণি → ঘুর্ণি
Run-on Error	ত্রিভুবন → ত্রিভুবনঅষ্টক
Split-word Error (Random)	মিহি → মি হি
Split-word Error (Left)	ঘোলাটে → ঘোলা টে
Split-word Error (Right)	অশান্তি → অ শান্তি
Split-word Error (Both)	শ্রেণিকক্ষ → শ্রেণি কক্ষ
Homonym Error	বর্ষা → বর্শা

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TABLE II
Errors by ERRANT Classification

Error Type	Example
Spelling	পরনির্ভরশীল → ফরনির্ভরশীল
Orthography	ব্যবসা প্রতিষ্ঠান → ব্যবসাপ্রতিষ্ঠান
Punctuation	। → !
Noun Inflection	অধিবাসীরা → অধিবাসী
Pronoun	আমি → আমরা
Verb Tense	যাবে → যায়
Adjective Form	মৃত (স্ত্রী) → মৃতা (স্ত্রী)
Subject-Verb Agreement	(সে) খায় → (সে) খাই
Conjunction	কিন্তু → এবং
Literary Register	পড়ে → পড়িয়া

- ▶ To detect sub-strings of a Bangla text that contain grammatical, punctuation, or spelling errors.

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For example:

Input	পুরা মাছ টাই খেলাম
Output	\$পুরা\$ \$মাছ টাই\$ খেলাম\$\$

- ▶ To detect sub-strings of a Bangla text that contain grammatical, punctuation, or spelling errors.

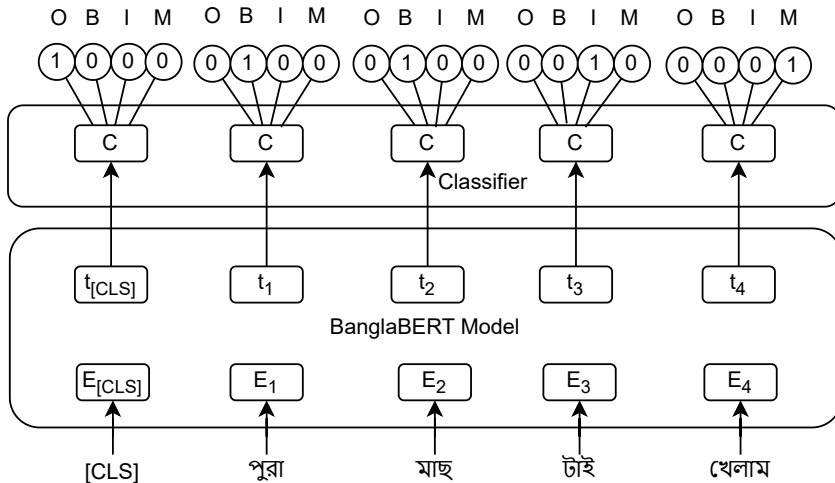
For example:

Input	পুরা মাছ টাই খেলাম
Output	\$পুরা\$ \$মাছ টাই\$ খেলাম\$\$

- ▶ Data
 - Training data: Around 20,000 texts with 7500 errors
 - Test data: 5,000 texts

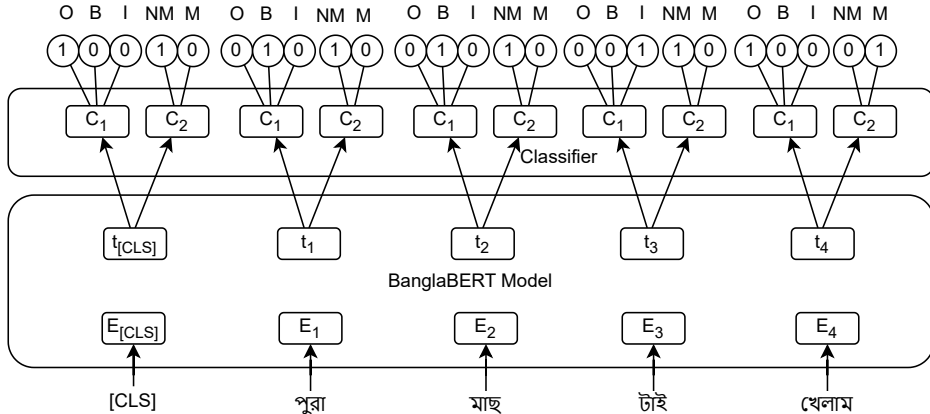
Token Classification Model I: 4 Classes

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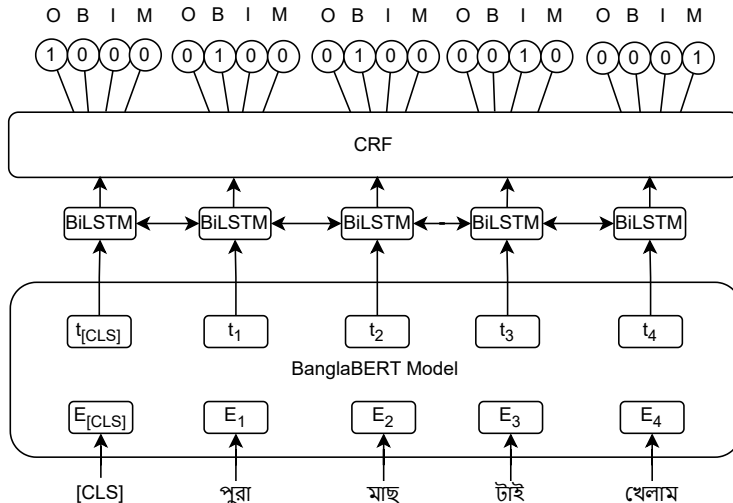
Token Classification Model II: 3+2 Classes

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Token Classification Model III: LSTM-CRF

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- ▶ We observe no performance gain
 - By adding BiLSTM and CRF
 - By modelling missing error separately

Table: Comparison of Token Class. Models on Dev Set

Model	Levenstein Distance
BanglaBERT-base 4 Classes	1.0239
BanglaBERT-base 3+2 Classes	1.0743
BanglaBERT-base+BiLSTM+CRF	1.0534

- ▶ We find BanglaBERT-base and BanglaBERT-large to perform best.

Table: Comparison of Transformers Models on Private Test Set

Model	Levenstein Distance
XLM-RoBERTa-base	1.3940
DeBERTa-V3-large	1.3552
BanglaBERT-base	1.2120
BanglaBERT-large	1.1844

- ▶ Mitigates overfitting and noise-modelling.
- ▶ 0.1 for BanglaBERT-base and 0.2 for BanglaBERT-large.

Table: Private test set results for label smoothing

Type	Levenstein Distance
BanglaBERT-large+standard CE	1.1640
BanglaBERT-large+smoothing factor 0.2	1.1588

- ▶ Unicode characters may have multiple representations

য় VS য়্‌

- ▶ De-normalized output using minimum edit distance alignment

Table: Results of Unicode normalization on the private test set

Type	Levenstein Distance
BanglaBERT-large without normalization	1.130
BanglaBERT-large with normalization	1.084

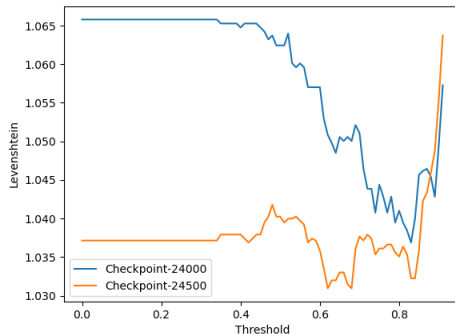
- ▶ We deterministically fix
 - Extra spaces before punctuation (space fix)
 - Missing punctuation at the end (end fix)

Model	Levenstein Distance
BanglaBERT-base	1.2948
BanglaBERT-base+space fix	1.246
BanglaBERT-base+space fix+end fix	1.212

► Steps

- Collect the database of common Bangla Spelling Error from DPCSpell Paper.
- Remove error in online Bangla dictionary.
- Remove Bangla Wikipedia title words.
- Finally, only apply if the word is not a named-entity as per the BNLP NER model.

► Result: Small gain (0.0008) in performance at Phase 1 test set.



BanglaBERT-large+threshold 0.0	1.1892
BanglaBERT-large+threshold 0.8	1.1588

- ▶ Type I
 - Union
 - Intersection
- ▶ Type II
 - Single-checkpoint
 - Three-checkpoints

Table: Effectiveness of Ensemble I on Private Test Set

Type	Levenstein Distance
BanglaBERT-large only	1.2212
BanglaBERT-base+large Union	1.2524
BanglaBERT-base+large Intersection	1.144

Table: Effectiveness of Ensemble II on Public Test Set

Model	Levenstein Distance
Single-checkpoint	1.0648
Three-checkpoints	1.0539

- ▶ Task formalization: Four-class Token Classification
- ▶ Models: BanglaBERT-base and BanglaBERT-large
- ▶ Ensemble:
 - Union
 - Three-checkpoints
- ▶ Loss function: Label Smoothing Cross-Entropy
- ▶ Optimizer: AdamW with Linear LR Scheduling
- ▶ Preprocessing: Normalization
- ▶ Postprocessing:
 - Confidence Thresholding
 - Denormalization
 - Rule-based Fix

- ▶ To employ self-training with in-domain unlabeled data.
- ▶ To combine self-training with feature-based learning to learn a more robust model.
- ▶ To use adversarial training strategies.