

HIERARCHICAL MULTI-LABEL NARRATIVE CLASSIFICATION ON ONLINE NEWS

Team 21

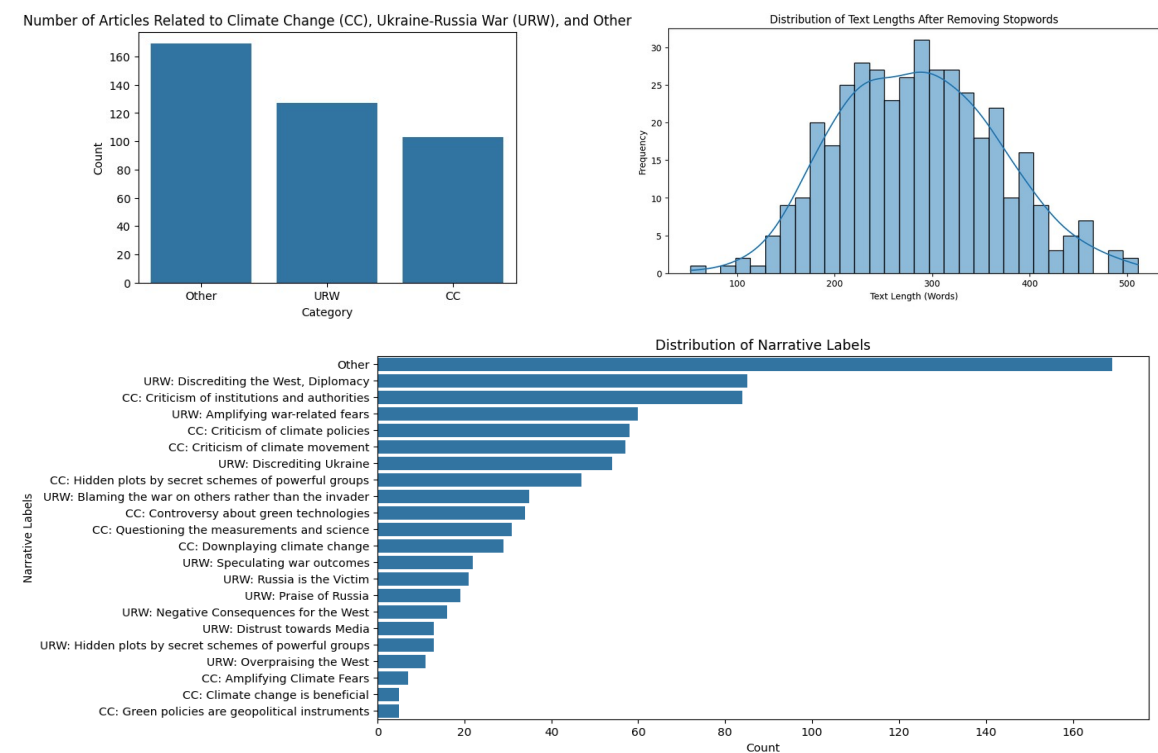
Prasun Dutta, Muhammad Zakria, Abrar Hassan, Muhammad Shazib, Wardah Noel

I. PROBLEM STATEMENT

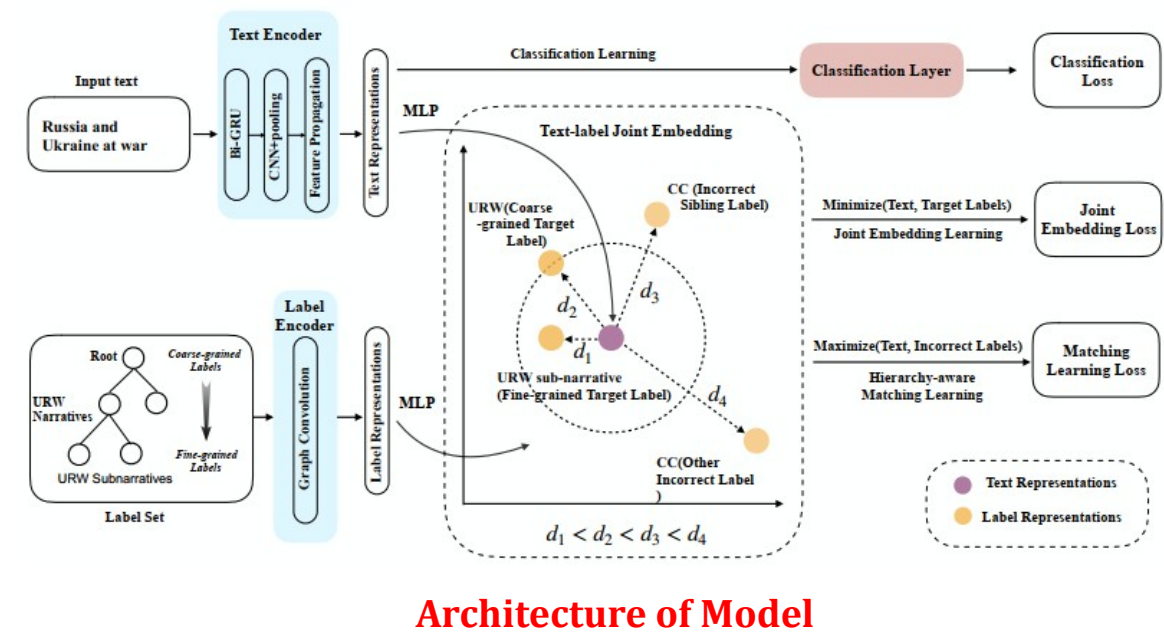
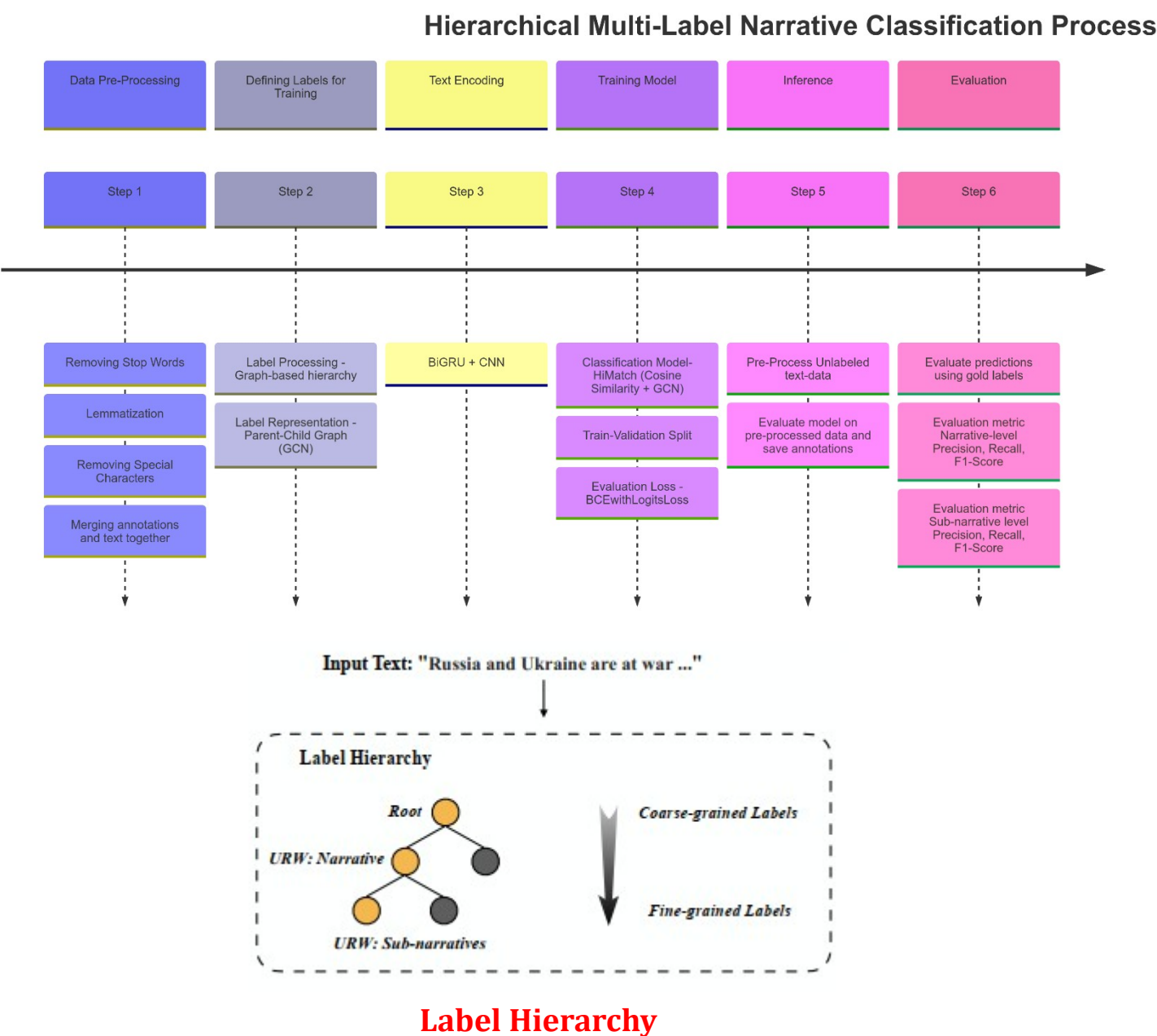
The goal is to automatically **classify narratives** from multilingual online news, focusing on **identifying** and **categorizing** narratives to mitigate disinformation and manipulation during major crisis events.

II. DATASET ANALYSIS

- The dataset contains three fields: **article_id**, **narrative labels**, and **sub-narrative labels**.
- Articles are assigned coarse-level narrative labels:
"CC": Climate Change
"URW": Ukraine-Russia War
and corresponding fine-level sub-narrative labels. If no sub-narrative fits, "Other" is used.



III. APPROACH



IV. RESULTS

- For model evaluation common evaluation metrics were used: **Precision**, **Recall**, and **F1-Score**. Both evaluated at a narrative and sub-narrative level.

Model Evaluation Results:
Narrative-Level Precision: 0.0623, Recall: 0.1610, F1-score: 0.0667
Sub-Narrative-Level Precision: 0.0780, Recall: 0.1383, F1-score: 0.0694

| Narrative Classification Report: | | | | |
|--|-----------|--------|----------|---------|
| | precision | recall | f1-score | support |
| CC: Climate change is beneficial | 0.00 | 0.00 | 0.00 | 1 |
| CC: Controversy about green technologies | 0.00 | 0.00 | 0.00 | 2 |
| CC: Criticism of climate movement | 0.00 | 0.00 | 0.00 | 8 |
| CC: Criticism of climate policies | 0.03 | 0.33 | 0.06 | 3 |
| CC: Criticism of institutions and authorities | 0.00 | 0.00 | 0.00 | 8 |
| CC: Downplaying climate change | 0.00 | 0.00 | 0.00 | 2 |
| CC: Green policies are geopolitical instruments | 0.00 | 0.00 | 0.00 | 3 |
| CC: Hidden plots by secret schemes of powerful groups | 0.09 | 0.50 | 0.15 | 4 |
| CC: Questioning the measurements and science | 0.33 | 0.50 | 0.40 | 4 |
| Other | 0.00 | 0.00 | 0.00 | 11 |
| URW: Amplifying war-related fears | 0.00 | 0.00 | 0.00 | 3 |
| URW: Blaming the war on others rather than the invader | 0.08 | 0.17 | 0.11 | 6 |
| URW: Discrediting Ukraine | 0.15 | 0.29 | 0.20 | 7 |
| URW: Discrediting the West, Diplomacy | 0.67 | 0.22 | 0.33 | 9 |
| URW: Distrust towards Media | 0.00 | 0.00 | 0.00 | 4 |
| URW: Negative Consequences for the West | 0.00 | 0.00 | 0.00 | 1 |
| URW: Overpraising the West | 0.00 | 0.00 | 0.00 | 1 |
| URW: Praise of Russia | 0.00 | 0.00 | 0.00 | 2 |
| URW: Russia is the Victim | 0.00 | 0.00 | 0.00 | 2 |
| URW: Speculating war outcomes | 0.00 | 0.00 | 0.00 | 4 |

Example Usage
text_sample = "climate change is something to worry about in future"

Narrative-Level Predictions:

- CC: Criticism of climate policies
- CC: Hidden plots by secret schemes of powerful groups

Sub-Narrative-Level Predictions:

- CC: Hidden plots by secret schemes of powerful groups: Blaming global elites
- CC: Criticism of climate policies: Climate policies have negative impact on the economy
- CC: Criticism of climate policies: Other
- CC: Hidden plots by secret schemes of powerful groups: Climate agenda has hidden motives

V. CONCLUSION & FUTURE WORK

- The poor evaluation metrics score is primarily due to training on very few amount of text datasets (399) as a result failed to learn the semantic meaning of text properly.
- In future, more focus will be given on Data Augmentation: to generate lot of synthetic data similar to training data so that the model can learn the semantics of the text more properly.
 - Can be done via generating text through paraphrasing and synonyms
 - Converting other language text files to English carefully
- Also, we can explore out other methods for other languages. This method was implemented on Hindi text files too but it failed to predict any results. Hence, a different approach was used for it (pre-trained BERT Transformer for hindi language)

REFERENCES

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