

CS5803: Natural Language Processing

Project Title: Disaster Tweet Classification

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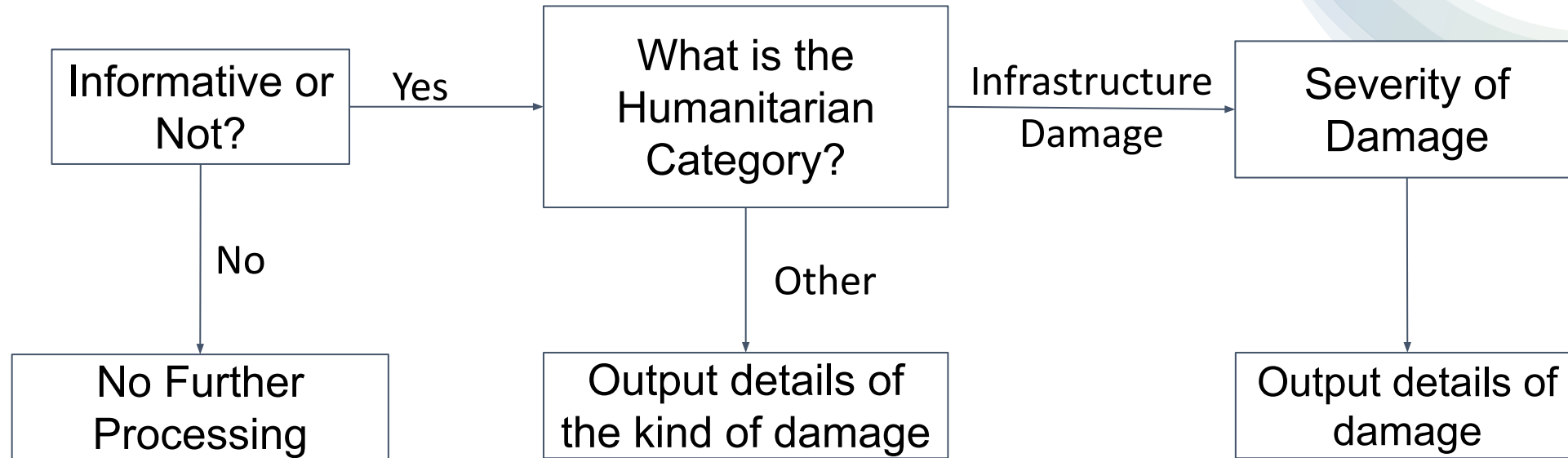
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Problem Statement

- During a disaster, people share huge information on Social Media which can be used to respond to the calamity.
- Proper and timely information can help the disaster response teams immensely.
- This project aims to classify tweets in informative and non-informative tweets to extract important information.
- The overall task involves three subtasks finding if a tweet contains information related to disaster, whether any humanitarian aid is required and how much damage has been caused.
- The task needs to be performed sequentially as a pipeline (if a task is informative then only it is processed further to see if humanitarian need is required).

Problem Statement



Dataset Details: CrisisMMD

- Several thousand manually annotated tweets and images.
- Tweets from seven major Natural Disaster worldwide : Hurricane Irma, Hurricane Harvey, Hurricane Maria, California Wildfires, Mexico Earthquake, Iran Earthquake, Sri Lanka Floods
- 3 type of annotation :-
 - Informative vs Not informative
 - Humanitarian categories
 - Damage severity assessment



Dataset Details: CrisisMMD

- Task 1: Informative vs Not informative
 - Informative
 - Not informative
- Task 2: Humanitarian categories
 - Affected individuals
 - Infrastructure and utility damage
 - Injured or dead people
 - Missing or found people
 - Rescue, volunteering or donation effort
 - Vehicle damage
 - Other relevant information
 - Non-Humanitarian
- Task 3: Damage severity assessment
 - Severe damage
 - Mild damage
 - Little or no damage
 - Don't know or can't judge

Dataset Details: CrisisMMD



Informative
Other relevant information



Informative
Affected individuals



Informative
Infrastructure and utility
damage
Severe damage



Not informative
Not humanitarian



Informative
Infrastructure and utility
damage
Severe damage



Informative
Infrastructure and utility
damage
Severe damage



Informative
Infrastructure and utility
damage
Severe damage



Informative
Infrastructure and utility
damage
Severe damage

Source: <https://aidr-dev2.qcri.org/apps/crisismmd>

Dataset Details: CrisisMMD

Crisis Name

[Hurricane Irma](#)

[Hurricane Harvey](#)

[Hurricane Maria](#)

[California wildfires](#)

[Mexico earthquake](#)

[Iraq-Iran earthquake](#)

[Sri Lanka floods](#)

Keywords Used to collect tweets

Hurricane Irma, Irma storm, Storm Irma, Irma Hurricane, Irma

Hurricane Harvey, Harvey, HurricaneHarvey, Tornado

Hurricane Maria, Maria Storm, Maria Cyclone, Maria Tornado, Tropical Storm Maria, HurricaneMaria, puerto rico

California fire, California wildfire, Wildfire California, USA Wildfire, California wildfires

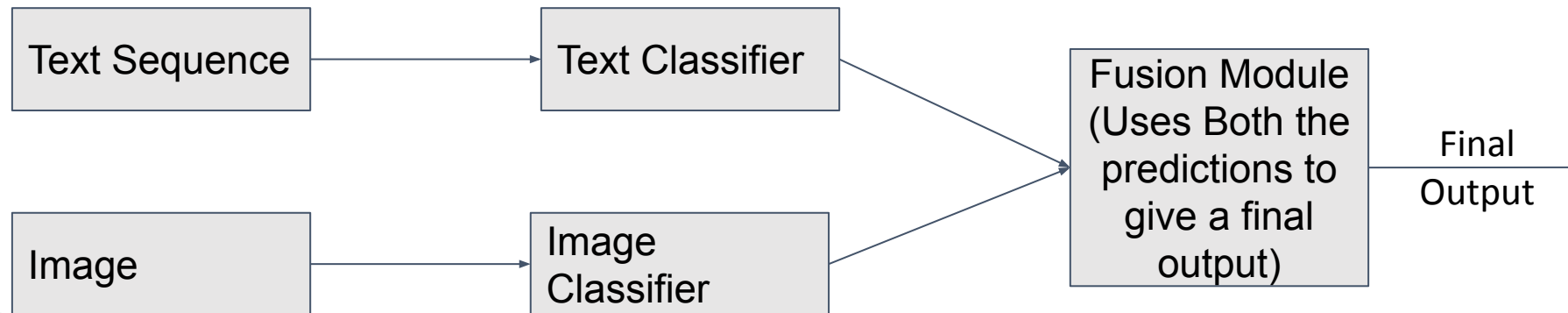
mexico earthquake, mexicoearthquake

kuwait earthquake, iran earthquake, halabja earthquake, Iraq earthquake

flood Sri Lanka, FloodSL, SriLanka flooding, SriLanka floods, SriLanka flood, typhoon mora, cyclone mora, mora, CycloneMora

Proposed Solution

- The problem is a multimodal (sequence and image) classification task.
- The labels of texts and images of the tweets are not always same and so the predictions from the two might also be different.



Proposed Solution

Fusion Module:

Various ways has been proposed in literature to fuse outputs in multimodal tasks, such as,

- Using an OR block
- Average the output probabilities from the two blocks
- Attention based decision fusion

We plan to implement these and see which one works best for this task.

Preliminary Results

- We have implemented a unimodal text classifier which classifies the tweet as informative or not.
- The implementation uses a bert language model followed by a linear classifier to classify the tweets.
- It sets a baseline and we will improve upon the results obtained in further work.

The results on validation dataset:

Accuracy : 0.73

f1 score : 0.83

Average training loss: 0.61

Accuracy: 0.72

F1 Score: 0.83

References

- Ferda Ofli, Firoj Alam, and Muhammad Imran, Analysis of Social Media Data using Multimodal Deep Learning for Disaster Response, In Proceedings of the 17th International Conference on Information Systems for Crisis Response and Management (ISCRAM), 2020
- Xukun Li, Doina Caragea, Improving Disaster-related Tweet Classification with a Multimodal Approach, Social Media for Disaster Response and Resilience Proceedings of the 17th ISCRAM Conference, 2020
- Hua XS., Zhang HJ., An Attention-Based Decision Fusion Scheme for Multimedia Information Retrieval. In: Aizawa K., Nakamura Y., Satoh S. (eds) Advances in Multimedia Information Processing - PCM 2004.

The image features a white background with decorative curved lines in the corners. In the top-right and bottom-left corners, there are thick, multi-layered curved lines that transition from a light blue color to a light green color. The text "Thank you." is centered in the middle of the page.

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