Documentation

Data Training:

- <u>Model Training:</u> The *LSTM* models were used to train using the data. The training process involved feeding the model with *input sequences* and their labels.
- **SMOTE: SMOTE** was further applied to the second training data to address **class imbalance**.

Findings (Metrics):

- <u>Classification report:</u> *Precision*, *recall*, *F1-score*, and *support* for each class were calculated to evaluate the model's performance.
- **Recall:** Reported for each class in the *classification report*.
- Class with least recall: The class with the *lowest recall* was identified to pinpoint areas for improvement.
- Confusion matrix: A visualization of the model's predictions versus actual labels, showing the number of true positives, true negatives, false positives, and false negatives for each class.
- Further analysis: Misclassified examples were examined for better understanding.
 Ratio of data points to total was also considered.

Data Cleaning:

- <u>Prediction on 'Others' class:</u> Titles initially classified as "*Others*" were further analyzed by model with *confidence score*.
- Confidence Threshold: A confidence threshold was applied. Predictions above 0.9 were accepted, below 0.6 were classified as "Others," and those between 0.6 and 0.9 were marked for manual review.

Accuracy, Recall:

- Accuracy: Reported on the test set. Model 1: 97.5%, Model 2: 93.2%
- Total recall (macro average recall): Calculated to provide a single score across all classes. Model 1: 93%, Model 2: 77%

Suggestions (mainly for the 2nd Model):

- <u>Hyperparameter Tuning:</u> Experimenting further with the *number of epochs*, *batch size*, and *learning rate*. *Early stopping* can also be used to avoid *overfitting*.
- <u>Data Augmentation:</u> Exploring techniques for *data augmentation* of the *underrepresented class* (the one with the *lowest recall*) to improve the model's performance on *minority classes*.
- <u>More Advanced Models:</u> Experimenting with more *advanced architectures* like *transformers* (*BERT*) or other deep learning models.
- **Ensemble Methods:** Combining predictions from multiple models can help reduce **bias** and **variance** and may lead to better generalization.