Technical Report

LegalEase: Summarization of legal documents and query-answering chatbot

30th March, 2024

Proposed Method

Issue with the old approach

1. **Legal-Pegasus**: We were initially using 8000+Indian judgemental data for fine tuning the model. We have the corresponding summaries dataset as well .But the issue was the tokenization limit of 1024. So, to increase the input capacity of the existing model, we started giving the whole document into chunks with the token limit of 1000. Meaning that the model will give a summary of every chunk. But this resulted in a huge summary output, not very useful.As well as,the accuracy score which was calculated using ROUGE score is not effective.

Updated Baseline Result : The F1 score of ROUGE-1 and ROUGE-L is around 0.2043.In new result the scores are updated to 0.3574 (for ROUGE-1) and 0.3294 (for ROUGE-L)

ROUGE-1 Precision: 0.1313
ROUGE-1 Recall: 0.7600
ROUGE-1 F1 Score: 0.2240

ROUGE-2 Precision: 0.0568
ROUGE-2 Recall: 0.4252
ROUGE-2 F1 Score: 0.1002

ROUGE-L Precision: 0.1198
ROUGE-L Recall: 0.6933
ROUGE-L F1 Score: 0.2043

ROUGE-1 Precision: 0.2526
ROUGE-1 Recall: 0.6618
ROUGE-1 F1 Score: 0.3574

ROUGE-2 Precision: 0.1236
ROUGE-2 Recall: 0.4049
ROUGE-2 F1 Score: 0.1836

ROUGE-L Precision: 0.2328
ROUGE-L Recall: 0.6112
ROUGE-L F1 Score: 0.3294

Old Result

Improved Result

In the above improved result ,We have followed the steps:

Breaking the judgment files into small chunks -> then using legal_pegasus and using mean_cosine similarity with the corresponding summary files -> generated summaries for each chunk -> training our model on these chunks and their corresponding summaries -> evaluating the results.

Methodology

To improve our model, we first need to train the model on an updated dataset.

1.Generation of our own dataset from existing data

We have 'Judgment'(d) and 'summary'(s) datasets. But as the input token is limited we cannot use the 'Judgment' dataset as a whole to train our model. As we are taking inputs in chunks, we cannot assume that 's' is the summary of all chunks. We need summaries for each chunk of input.

Approach:

- 1. Suppose we have 'd' as a complete document and 's' is its summary. Using Mean cosine similarity, we first map every sentence in 's' to the most similar sentences in 'd'.
- 2. Now, initially we have input in chunks format, let say {d1,d2,...dn} making a total n chunks.So, for every chunk 'di', we do Mean cosine similarity with 's',so to get important/relevant sentences in 'di', which will contribute to the summary. Combining all these important sentences, we get the summary 'si' for the chunk 'di'.
- 3. In this process some chunks have a lot of important sentences, while some don't have any.
- 4. Now, for each document in the training data, we have chunks of datasets and their corresponding summaries generated by Mean cosine similarity. So, our training dataset is now suitable, which will help in training the model better. As our model takes input in chunks and generates summaries.

The output we generated:

Here we see the summaries for each chunk. We have stored all the chunks and corresponding summaries in specified files.

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It raises the question whether munici pal property tax and urban immoveable property tax payable under the relevant Bombay Acts are allowable deductions under section 9 (1) (iv) of the 1.

The court held that the income from the property was not subject to the tax provisions of Section 9 (1) of the Income Tax Act and therefore the tax officer had no power to determine the The Incibural, however, agreed to refer two questions of law to the High Court of Judicature at Bombay, namely, (1) Whether the municipal taxes paid by .cn the applicant company are an all Chunk 3: Summary - The question for our determination is whether the municipal property tax and urban immoveable property tax and be deducted as an allowance under clause (uv) of sub st The court held that the tax shall be payable by an assessee under the head 'income from property' in respect of the bona fide annual value of property constructing of any buildings or lar Chunk 4: Summary - The Gawman's Act of 1939 amended the tax treatment of interest paid on the security of a property acquired, constructed, repaired, renewed or reconstructed with borro. Chunk 6: Summary - The Court of Appeal has ruled that the principle that the outgoings were to part of the assessee's income at all should be extended to cases where obligatory payment. The Court also held that, in view of the opening words of the new sub clause in the Act, the expression "capital change" also cannot have reference to a charge on the property, and we ti Chunk 7: Summary - The sub clause provides for a deduction of annual sums so charged, such sums not being capital sums, the limiting words being intended to exclude cases where capital x Cannal 3., however, took a different view and observed that he was not prepared to accept the sugges tion that a document which provides for a certain payment to be made monthly or annual Chunk 9: Summary - Section 143 of the City of Bombay Municipal Act, 1888, authorises the levy of a general tax

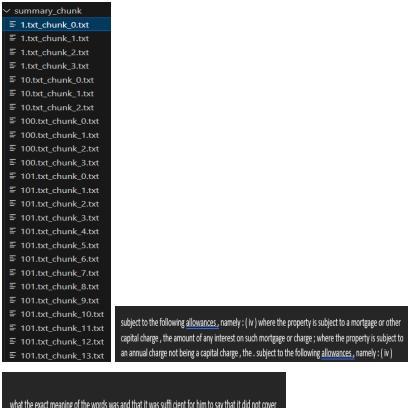
The stored Chunk_Judgement:

judgment_chunk ≡ 1.txt_chunk_0.txt ≡ 1.txt chunk 2.txt ■ 1.txt chunk 3.txt ■ 10.txt chunk 0.txt ≡ 10.txt_chunk_2.txt ≡ 100.txt chunk 0.txt ≡ 100.txt_chunk_1.txt ≡ 100.txt chunk 3.txt ≡ 101.txt_chunk_0.txt ≡ 101.txt_chunk_2.txt ≡ 101.txt chunk 3.txt ≡ 101.txt chunk 5.txt ≡ 101.txt chunk 6.txt ≡ 101.txt chunk 7.txt ≡ 101.txt_chunk_8.txt ≡ 101.txt_chunk_10.txt ≡ 101.txt chunk 13.txt ≡ 101.txt chunk 14.txt ≡ 101.txt chunk 15.txt ≡ 101.txt chunk 16.txt

Appeal No. LOVI of 1949. Appeal from the High Court of judicature, Bombay, in a reference under section 66 of the Indian Income tax Act, 1022. E.M., Murahi (IA. P. Nathone), with him), for the appoint and the Indian Income tax Act, 1022. E.M., Murahi (IA. P. Nathone), with him), for the appoint and income tax Act, 1022. E.M., Murahi (IA. P. Nathone), with him), for the appointed responsibility of the Court was delinered by MERR CHAND MAHAIAN L. This is an appeal against a judgment of the High Court of Judicature at Bombay in an income tax matter and trains to equation whether municipal property tax and urban immovable property tax apable under the relevant Bombay Acts are allowable deductions under section 9 (1.5) (by of the Indian Income tax Act. The assessment was not appeal and the training that the court of the appeals of the term of the assessment was a long to the term of the assessment was investment company deriving its income from properties in the city of Rambay. For the assessment was 1940 4th the net income of the assessment was recommended to the control of the Act and the Act and

and so should not be included in his assessment. Though the decision proceeded on the principle that the outgoings were not part of the assesses 'is known at all, the frames of the amending Act of 1333 wanted, apparently, to extend the principle, so far as the assessment of property was concerned, even to case where obligatory ayaments had to be made on of the asposses' is known from the reporting classification and the second sub clauss, namely, "where the property is subject to an amrual charge not being a capital charge, the amount of such charge 'was older, it is this sub clause which the appellatin revokes in support of its claim to deviction of the municipal and intons, property tases in the present case. In view of the opening words of the newly added sub clause, the expression 'critical charge' all outset therein anoth have reference to a charge on the property, and we think it must (1) LLR. 60 cal. 557 be understood in the same sense as in sub clause [1]; that is to say, the first sub clause having provided for deeds, togo of interest where a capital sum is charged on the property, this sub clause provides for a deduction of annual sums so charged, such sums not being capital sums, the limiting words being intended to exclude cases where capital reade on the security of the property, this sub clause hands provided for a deduction of annual sums so charged, such sums not being capital sums, as lending the sums of the property in the property in the sub-clause of the property in the pro

Chunk Summary:



what the exact meaning of the words was and that it was sufficient for him to say that it did not cover municipal taxes which are made a charge on the property under section 212 of the Bombay Municipal

2. Fine-tuning our existing model on this updated dataset.

As we take input in chunks, our model will predict the most essential statements in each chunk and then combine all these outputs to give a final summary. This method will also reduce the summary length compared to the old output. As some chunks may not have any important statements.

Finally, we are doing this:

1. **Preparation**: Importing necessary libraries.

Reading judgment files.

2. Chunking and Summarizing:

Breaking down judgments into smaller chunks.

Generating summaries for each chunk using Mean Cosine similarity between the generated chunk and actual summary we have.

3. **Fine-Tuning** LegalPegasus on the new data set we generated.

Train LegalPegasus with generated summaries, which is chunk wise summarize.

4. Evaluation:

Assessing LegalPegasus performance on test data.

5. Analysis:

Analyzing results for further refinement

Evaluation

Our evaluation approach includes

- For evaluation, we have used 3 variants of ROUGE, that is (ROUGE-1,2 and L) scores as it is best for Natural languages.
- We have used Divide and conquer techniques to deal with large datasets.
- Compared our accuracy with the accuracy given by the same model in the below research paper.
- Fine Tuning our model on a huge dataset, for better performance. Even though we were able to generate a good amount of chunk data and their summaries, we still need to increase the training dataset.

Conclusion

Legal Pegasus was already fine-tuned on legal documents. We improved the input size and used Mean cosine similarity function for generating our own dataset and their summaries. We will train our model using these chunks and summary pairs . We hope to achieve a F1 score of atleast 0.5 and then again proceed further.

The research paper we are following has an F1 score of around 0.864. So, we will try to achieve that level of F1 score.

References:

https://arxiv.org/abs/2004.06190

https://aclanthology.org/2022.aacl-main.77.pdf