

LTRC@MuP-2022: Multi-Perspective Scientific Document Summarization Using Pre-trained Generation Models

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Introduction

MuP2022 shared task

- Scientific
 documents with
 multiple summaries
- Leveraging multiple
 gold summaries to
 generate one
 multi-perspective
 summary.
- Explored several pretrained models such as BART, T5,
 ProphetNet.
- Two Stage
 Fine-tuning
 approach: First
 finetuning BART_{large}
 on SciTLDR dataset
 and then on the MuP
 dataset.
- We secured 3rd rank
 in the MuP-2022
 shared task.

Fine Tuning Pre Trained Models

Parameters	BART	T5	ProphetNet		
Max source length	1024	1024	512		
Max target length	150	128	128		
Min target length	56	30	56		
Batch Size	1	1	1		
Epochs	2	10	1		
Vocab Size	50265	32128	30522		
Beam Size	4	4	5		
Learning Rate	5e-5	1e-4	5e-5		

Model	R-1 R-2		R-L	Avg R-f		
Baseline	40.8	12.3	24.5	25.8		
BART _{large} cnn	40.68	12.47	24.99	26.05		
DistilBART cnn	39.36	11.79	24.47	25.21		
$BART_{base}$ cnn	39.12	11.42	23.8	24.78		
$T5_{base}$	38.35	11.26	24.64	24.75		
ProphetNet	38.15	11.45	24.25	24.62		
BART_{base}	38.53	11.39	23.92	24.61		
ProphetNet cnn	37.59	10.91	24.09	24.2		
DANCER + BART	33.07	9.06	18.2	20.11		
BART + Two-stage	32.51	6.82	20.64	19.99		

Experimenting with Input Data

To investigate the contribution of various sections of the scientific documents in the target summaries, we created the following four categories of source content:

- 1. Introduction
- 2. Abstract + Introduction
- 3. Abstract + Introduction
 - + Conclusion
- 4. Abstract + Conclusion

					Train & Val Data			Test Data			2	
	R-1	R-2	R-L	Avg R-f	1	2	3	4	1	2	3	4
Г	40.68	12.47	24.99	26.05	~				2	✓		
	40.67	12.5	24.93	26.03	~						/	
	40.47	12.29	24.76	25.84			/			/		
	40.34	12.28	24.79	25.8				/				/
	40.33	12.28	24.75	25.79		/				/		
	40.39	12.25	24.73	25.79			/				/	
	40.23	12.32	24.77	25.77	✓							~
	40.23	12.17	24.6	25.67				/		/		
	40.1	12.25	24.63	25.66		/			~			
	40.22	12.13	24.54	25.63	/				/			

We fine-tuned and tested BART_{large}cnn with all combinations of the above categories of data (consuming one as train and the other as test set). We found that, among these combinations, the model trained with **Introduction** and tested with **Abstract + Introduction** performed the best.

Analysis & Conclusion

- Among all, the BART cnn model performed the best.
- While this task considers summaries from multiple reviewers as different "perspectives", most of these summaries cover only the major contributions of the paper.
- These summaries, though diverse in their construction, do not look at the research paper from different points-of-view.
- We see a validation of this claim in our experiments where a model trained on "introduction" section alone outperforms all other combinations.