# AnyRef - Document Summarisation and Its Evaluation Using Metrics

Presented By: Lohith Reddy Kalluru, Md Zahid Hasan

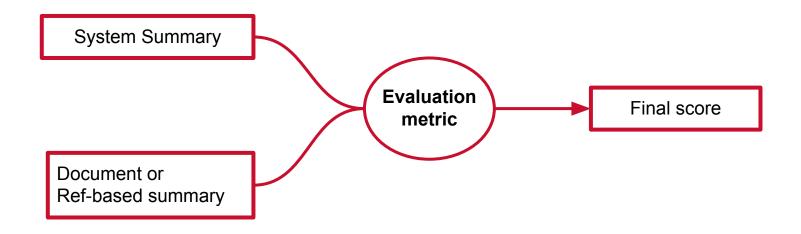
### **Outline**

Models (summarizer) Summarizer Datasets (Newsroom) Pseudo\_ref function **Evaluation Metric EvalBase framework ROUGE** score Scores Results Generalization

#### Introduction

- The method of extracting these summaries from the original huge text without losing vital information is called as Text Summarization. It is essential for the summary to be a fluent, continuous and depict the significant.
- The approach is to identify the important sections, interpret the context and reproduce in a new way. This ensures that the core information is conveyed through shortest text possible.
   Note that here, the sentences in summary are generated, not just extracted from original text.

## **Summarization evaluation/metrics**



#### **Dataset**

- Currently, we are using newsroom eval dataset for testing purposes.
- There are 60 articles, 7 systems, and 3 ratings by different people for each system for a total of 1,260 rows. Each row is a single rating across the four dimensions from the paper. The columns are:
  - 1. System name of the sumarization system/baseline being rated
  - 2. ArticleID index of the article in the dataset, can be used to group by article
  - 3. ArticleText, ArticleTitle, SystemSummary three inputs the rater sees (HTML encoded for AMT)
  - 4. CoherenceRating, FluencyRating, InformativenessRating, RelevanceRating ratings (1-5)

											.53
2140	fragments	A worker sets up a polling station the morning	John Avlon voter turnout in the is a sign of a	Why has GOP turnout taken a dive?	;	2	3	2	2 3	John Avlon voter turn	.50
2140	textrank	A worker sets up a polling station the morning	In New Hampshire , the same dynamic applied	Why has GOP turnout taken a dive?	,	4	5	4	5	John Avlon voter turn	

ArticleTitle CoherenceRating FluencyRating InformativenessRating RelevanceRating

2

4

ArticleText

A worker sets up

a polling station

A worker sets up

a polling station

the morning...

the morning...

SystemSummary

John Avlon voter

turnout in the is a

John Avlon voter

turnout in the is a

sign of a...

sign of a...

Why has GOP

turnout taken a

Why has GOP

turnout taken a

dive?

dive?

ArticleID

System

2140 fragments

2140 fragments

ReferenceSummary

John Avlon says low

voter turnout in the

John Avlon says low

voter turnout in the

prima...

prima...

## pseudo\_func

- → Input
  - documents
  - system\_summaries
  - summarizer name
  - ref\_based\_metric\_name
- → Output
  - scores

summarizer\_name: Hugging face model Hub

ref\_metric: [ROUGE, BERTscore]

```
def pseudo_func(predictions, references, model_name, ref_based_metric_name):
    device = "cuda" if torch.cuda.is_available() else "cpu"
    tokenizer = AutoTokenizer.from_pretrained(model_name)

# https://stackoverflow.com/questions/70544129/transformers-asking-to-pad-but-the-tokenizer-does-not-have-a-padding-token
    if tokenizer.pad_token is None:
        tokenizer.add_special_tokens({'pad_token': '[PAD]'})

if "google" in model_name.lower():
    model = PegasusForConditionalGeneration.from_pretrained(model_name).to(device)
    else :
    model = AutoModelForCausalLM.from_pretrained(model_name).to(device)
```

## **EvalBase/env.py**

#### Takes documents and extracts

- ArticleText
- SystemSummary
- ReferenceSummary
- hum\_eval\_path
- refs\_path
- hum\_with\_ref\_path

```
datasets = {
   "newsroom": {
        "human_metrics": ["InformativenessRating", "RelevanceRating", "CoherenceRating", "FluencyRating"],
        "docID_column": "ArticleID",
        "document_column": "ArticleText",
        "system_summary_column": "SystemSummary",
        "reference_summary_column": "ReferenceSummary",
        "approaches": ["trad", "new"],
        "human_eval_only_path": "dataloader/newsroom-human-eval.csv", # you need to get this file. See ReadMe.
        "refs_path": "dataloader/test.jsonl", # you need to get this file. See ReadMe.
        "human_eval_w_refs_path": "dataloader/newsroom_human_eval_with_refs.csv"
},
```

https://github.com/SigmaWe/EvalBase/blob/main/env.py

## Test framework (EvalBase)

- We plugged in the functions through env.py which are partial functions with certain arguments sent to the eval\_utils.py when called from newsroom.py.
- Datasets were read automatically without the need for any change.
- It can evaluate through multiple metrics and aggregate the results in json and text format.

```
"google-pegasus-xsum":functools.partial(pseudo_func.pseudo_func,model_name = "google/pegasus-xsum", ref_based_metric_name= "rouge"),
"bert-base-cased":functools.partial(pseudo_func.pseudo_func,model_name = "bert-base-cased", ref_based_metric_name= "rouge"),
"distilgpt2":functools.partial(pseudo_func.pseudo_func,model_name = "distilgpt2", ref_based_metric_name= "rouge"),
"openai-gpt":functools.partial(pseudo_func.pseudo_func,model_name = "openai-gpt", ref_based_metric_name= "rouge"),
"google-pegasus-cnn_dailymail":functools.partial(pseudo_func.pseudo_func,model_name = "google/pegasus-cnn_dailymail",ref_based_metric_name= "rouge"),
"gpt2":functools.partial(pseudo_func.pseudo_func,model_name = "gpt2", ref_based_metric_name= "rouge")
```

## **Initial Results (Dec 8)**

Model= Pegasus and Eval\_metric= ROUGE:

Rogue-1	Rogue-2	RougeL	RougeLsum
0.04743833017077	0.03802281368821	0.04174573055028	0.04174573055028
0.04743833017077	0.03802281368821	0.04174573055028	0.04174573055028
0.08379888268156	0.08208955223880	0.08379888268156	0.08379888268156

## **Initial Results (Dec 8)**

Model= bert-base-uncased, Eval\_metric= ROUGE for first 15 newsroom documents:

	Rogue-1	Rogue-2	RougeL	RougeLsum
BERT/ROUGE	0.34360992309416	0.3031325629862	0.33172716963627	0.32889254151974

Model= bert-base-uncased, Eval\_metric=BERTScore for first 15 newsroom documents:

	avg_precison	avg_recall	avg_F1-score
BERT/ROUGE	0.83231127262115	0.89182925224304	0.8610429709

## **Sample of final Results (Dec 17)**

corr metric	aspect	approach	model	score_name	
pearsonr	InformativenessRating	new	google-pegasus-xsum	rouge1	0.369
	3		3 3 1 3	rouge2	0.336
				rougeL	0.295
				rougeLsum	0.295
			bert-base-cased	rouge1	0.764
				rouge2	0.716
				rougeL	0.740
				rougeLsum	0.740
			distilgpt2	rouge1	0.776
				rouge2	0.766
				rougeL	0.773
				rougeLsum	0.773
			openai-gpt	rouge1	0.766
				rouge2	0.719
				rougeL	0.741
				rougeLsum	0.741
			pegasus-cnn_dailymail	rouge1	0.638
				rouge2	0.536
				rougeL	0.572
				rougeLsum	0.572
			gpt2	rouge1	0.776
				rouge2	0.766
				rougeL	0.773
				rougeLsum	0.773

Full result folder in the Git repository: <a href="https://github.com/SigmaWe/AnyRef">https://github.com/SigmaWe/AnyRef</a> team 1/tree/main/results

#### **Future directions**

- Extend the function to accept other arguments related to summariser model or metric → generalize
- Other downstream tasks and comparison

## THANK YOU