



PRIMERA

Pyramid-based Masked Sentence Pre-training for Multi-document Summarization

Wen Xiao¹, Iz Beltagy², Giuseppe Carenini¹, Arman Cohan²

¹University of British Columbia

²Allen Institute of Artificial Intelligence

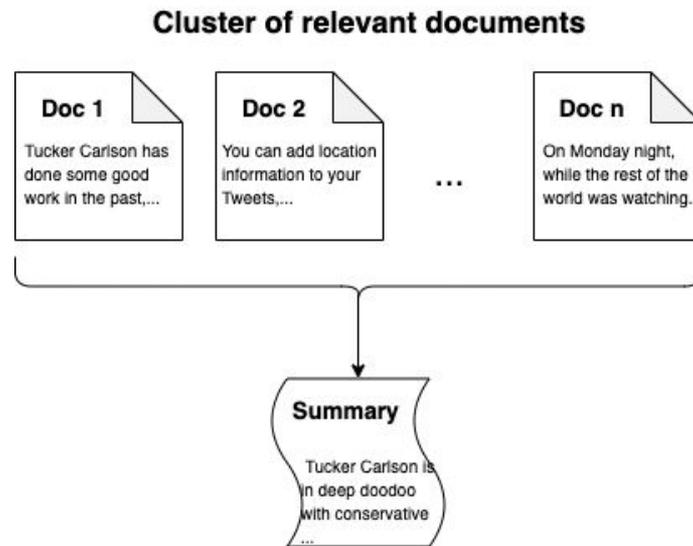
What is Multi-Document Summarization (MDS)?

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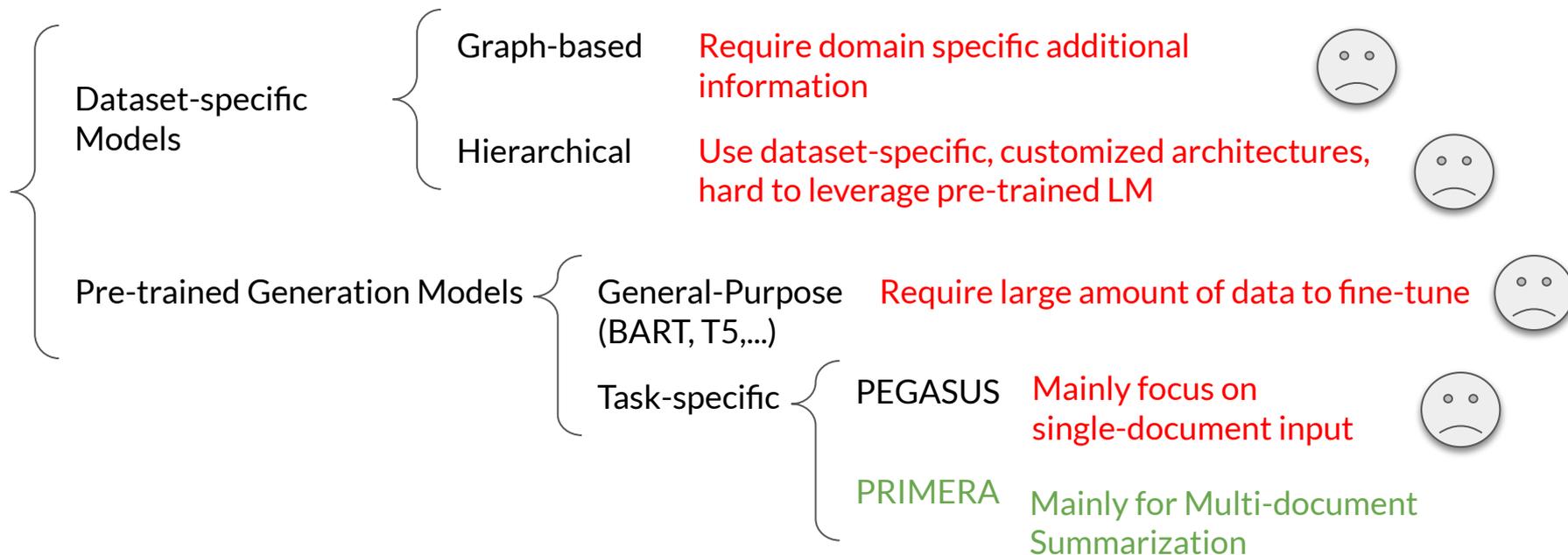
Task: Generate a summary given a **cluster of relevant documents**,

e.g.

- News articles
- Scientific papers

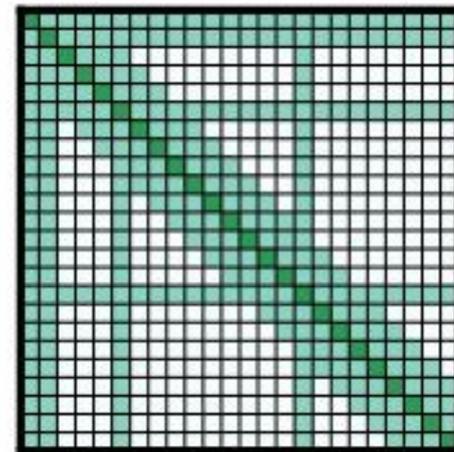


Previous Methods for MDS



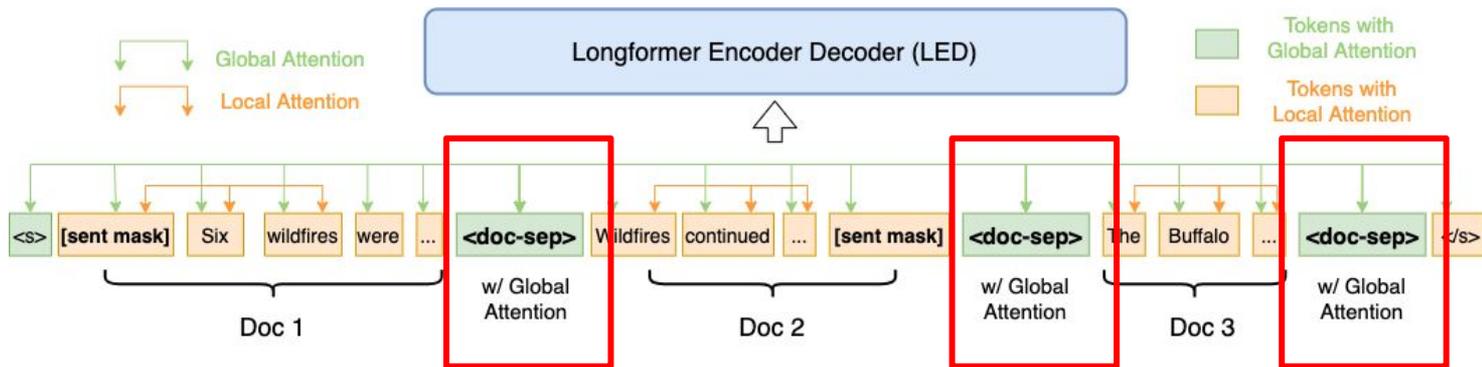
Overview of PRIMERA

- Architecture: Longformer Encoder Decoder (LED)
 - Global + Local Attention
 - Allows for long sequence inputs

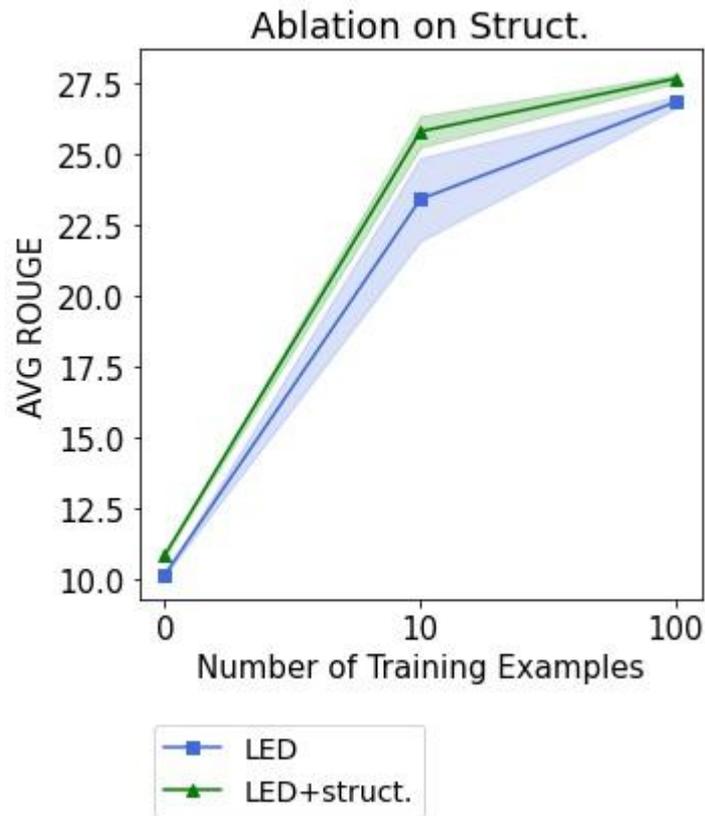


Overview of PRIMERA

- Architecture: Longformer Encoder Decoder (LED)
 - Global + Local Attention
 - Allows for long sequence input
- Input Structure:
 - documents separated with document separator(<doc-sep>)
 - Global Attention on <doc-sep>



Impact of the Proposed Input Structure



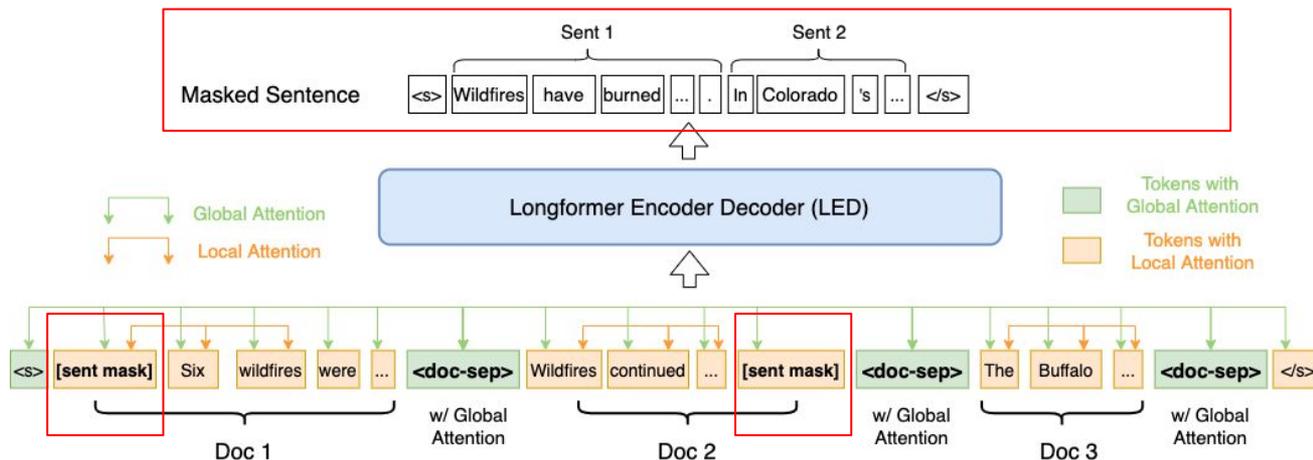
- Input structure improves the results

Overview of PRIMERA

- Architecture: Longformer (with local and global attention)
 - Global + Local Attention
 - Allows for long sequence input
- Input Structure:
 - documents separated with document separator(<doc-sep>)
 - Global Attention on <doc-sep>
- Pre-training:
 - **Goal:** Teach the model to identify and aggregate salient information across a “cluster” of related documents
 - **Multi-doc corpus:** Newshead (360k clusters, 3.5 doc/cluster on average)
 - **Objective:** Gap Sentence Generation (GSG)
 - **Novel Masking Strategy:** Entity Pyramid

Pre-training : Objective

- **Gap Sentence Generation [Zhang et al., 2020]:**
 - Select several **SALIENT** sentences from the input documents (as pseudo-summary)
 - Mask out the selected sentences
 - Generate them in order in the decoder



How to select **SALIENT** sentences?

Previous work for single document input (PEGASUS):

- Random
- Lead-K
- Principle **Best**
 - **Intuition:** select the **most central** sentences in the document
 - The score is defined as the ROUGE score between **each sentence** and **rest of the document**

$$\text{Score}(s_i) = \text{Rouge}(s_i, D / \{s_i\})$$

How to select **SALIENT** sentences?

- However, multi-document input tends to be more **redundant** than single document input.
- And such strategy would prefer **exact match between sentences**, resulting in selection of less representative information.

Example of the problem with vanilla SGS

Doc #1

Wildfires have burned across tens of thousands of acres of parched terrain in Colorado, spurring thousands of evacuations ..., residents have sought shelter in middle schools, and local officials fear tourists usually drawn to the region for the summer may not come.

Doc #2

... In Colorado's southwest, authorities have shuttered the San Juan National Forest in southwestern Colorado and residents of more than 2,000 homes were forced to evacuate. No homes had been destroyed ... "Under current conditions, one abandoned campfire or spark could cause a catastrophic wildfire, ..., with human life and property," said San Juan National Forest Fire Staff Officer Richard Bustamante...

Doc #3

The Buffalo Fire west of Denver is ... Several wildfires in Colorado have prompted thousands of home evacuations ... Nearly 1,400 homes have been evacuated in Summit County, Colorado, "Under current conditions, one abandoned campfire or spark could cause a catastrophic wildfire, ... , with human life and property," said Richard Bustamante, SJNF forest fire staff officer ...

New Masking Strategy: Entity Pyramid

Goal:

Select sentences that best represent the entire cluster of input documents



Pyramid Evaluation

with multiple refs

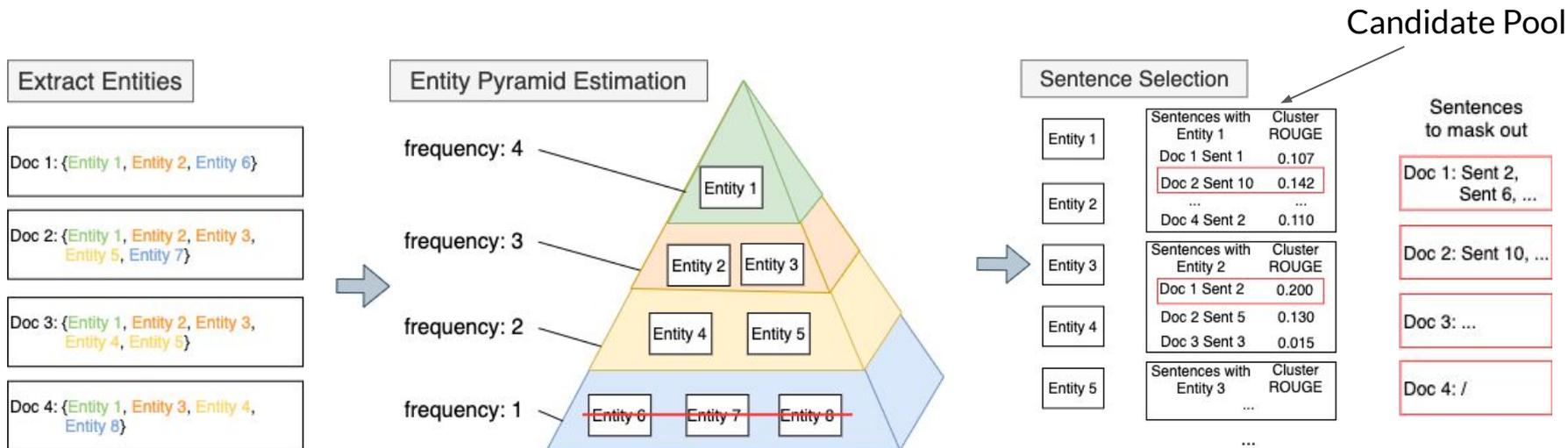
- The importance of information is quantified by the frequency of the **gold references** that include it.
- The **more gold references** include a fact, the more important it is.
- The facts are identified by **human-labeled Summary Content Units (SCUs)**

Entity Pyramid

with multiple docs

- The importance of information is quantified by the frequency of **the documents** that include it.
- The **more documents** include a fact, the more important it is.
- The facts are identified by **Entities**

New Masking Strategy: Entity Pyramid



Cluster ROUGE: $Score(s_i) = \sum_{\{doc_j \in C, s_i \notin doc_j\}} ROUGE(s_i, doc_j)$

Example

Doc #1

Wildfires have burned across tens of thousands of acres of parched terrain in **Colorado**, spurring thousands of evacuations (0.107) ..., residents have sought shelter in middle schools, and local officials fear tourists usually drawn to the region for the summer may not come.

Doc #2

... In Colorado's southwest, authorities have shuttered the San Juan National Forest in southwestern Colorado and residents of more than 2,000 homes were forced to evacuate. (0.187) No homes had been destroyed ... "Under current conditions, one abandoned campfire or spark could cause a catastrophic wildfire, ..., with human life and property," said San Juan National Forest Fire Staff Officer Richard Bustamante...

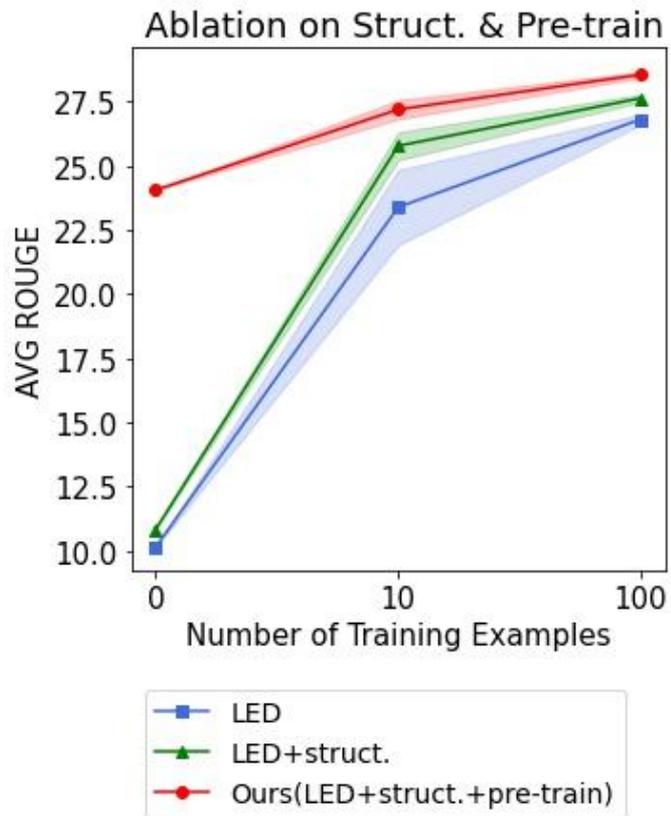
Doc #3

The Buffalo Fire west of Denver is ... Several wildfires in **Colorado** have prompted thousands of home evacuations (0.172)... Nearly 1,400 homes have been evacuated in Summit County, **Colorado**, (0.179)..... "Under current conditions, one abandoned campfire or spark could cause a catastrophic wildfire, ... , with human life and property," said Richard Bustamante, SJNF forest fire staff officer ...

Entity List

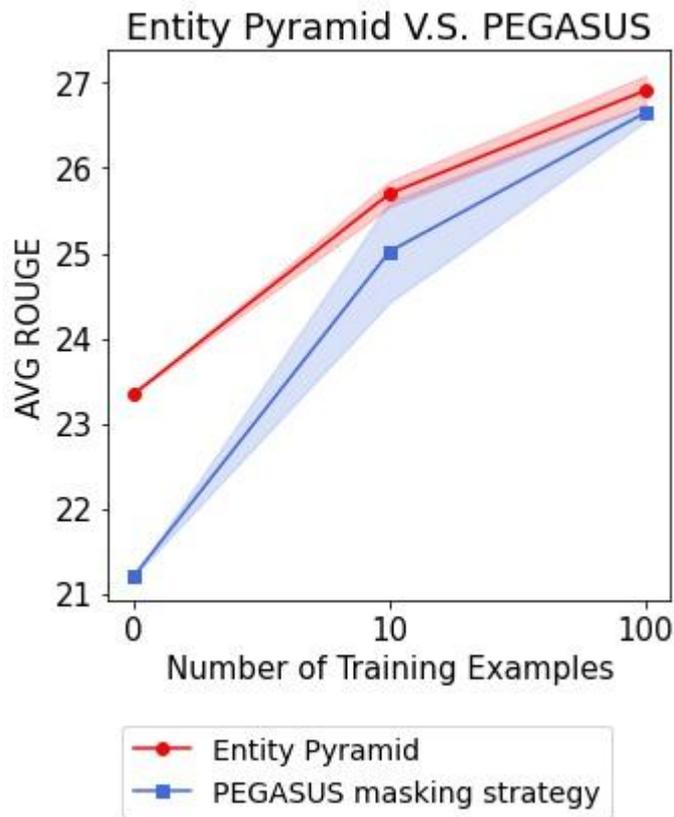
Colorado(3), **Wildfires(3)**, **416(2)**, **Tuesday(2)**, **San Juan National Forest(2)**,....

Impact of Proposed Pre-training



- Pre-training helps improving the model, especially for the zero-shot setting.

Impact of Pre-training Strategies

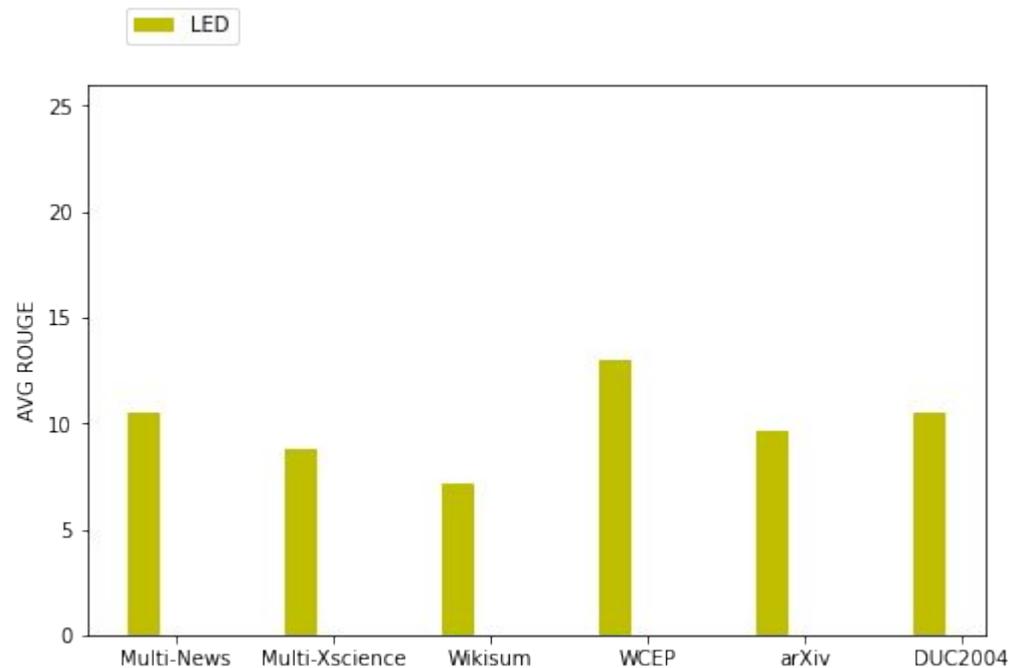


- Same architecture (LED-Base)
- Same input structure
- Same pre-training objective
- Same pre-training dataset
- Zero/Few-shot setting
- **The Entity Pyramid strategy works better than the Principle strategy used in PEGASUS.**

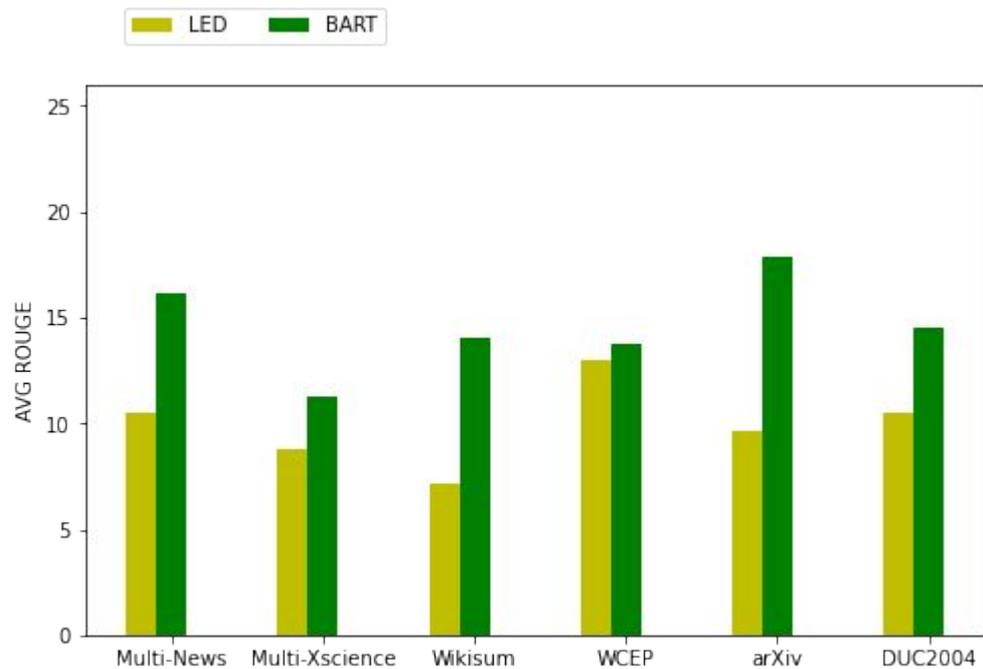
Experiments - Automatic Evaluation

- Evaluation Datasets:
 - Multi-Doc.: Multi-News, Multi-XScience, WCEP, Wikisum, DUC2004
 - Single Doc. arXiv
- Settings:
 - Zero-shot (with length limit)
 - Few-shot: 10/100 training examples, 5 runs for each model
 - Fully supervised
- Compared Models:
 - BART
 - PEGASUS
 - Longformer Encoder Decoder (LED)
 - Prior SOTA Models (fully supervised only)
- Evaluation Metric:
 - ROUGE scores (R-1, R-2, and R-L)

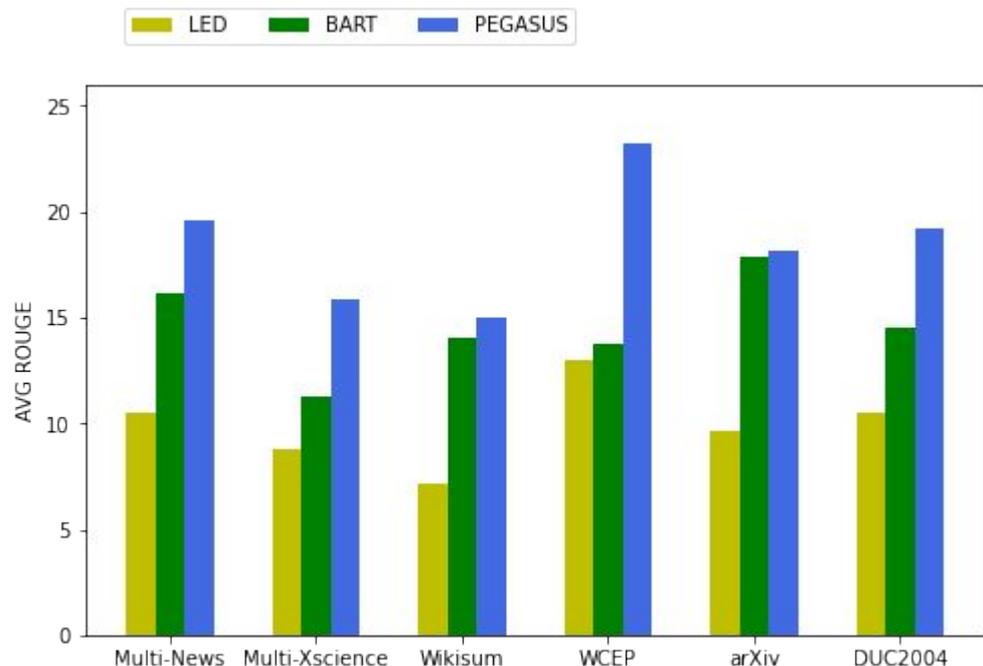
Results on Zero-shot



Results on Zero-shot

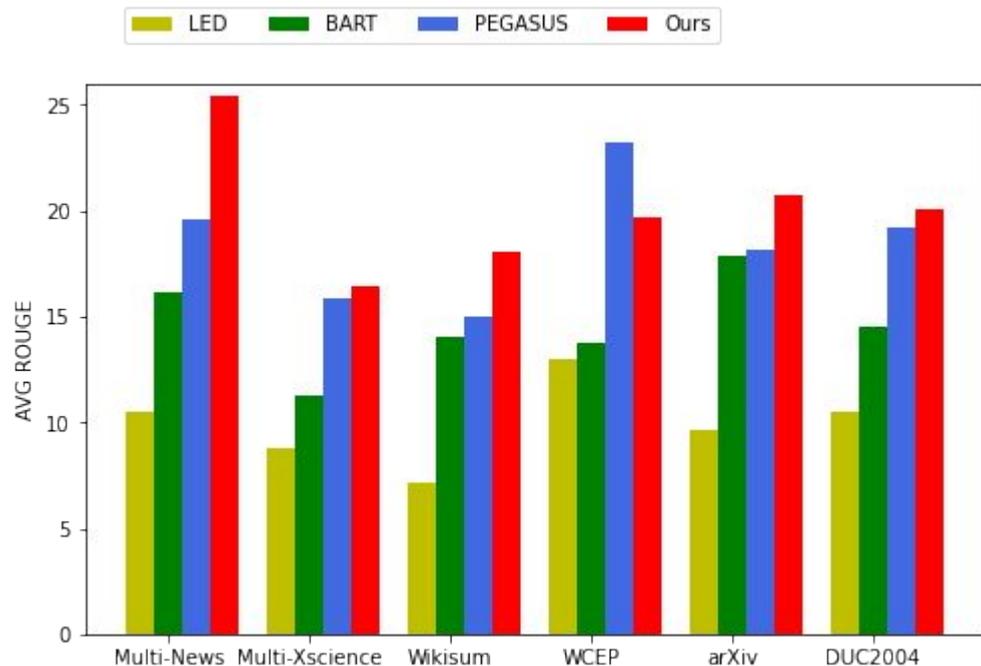


Results on Zero-shot



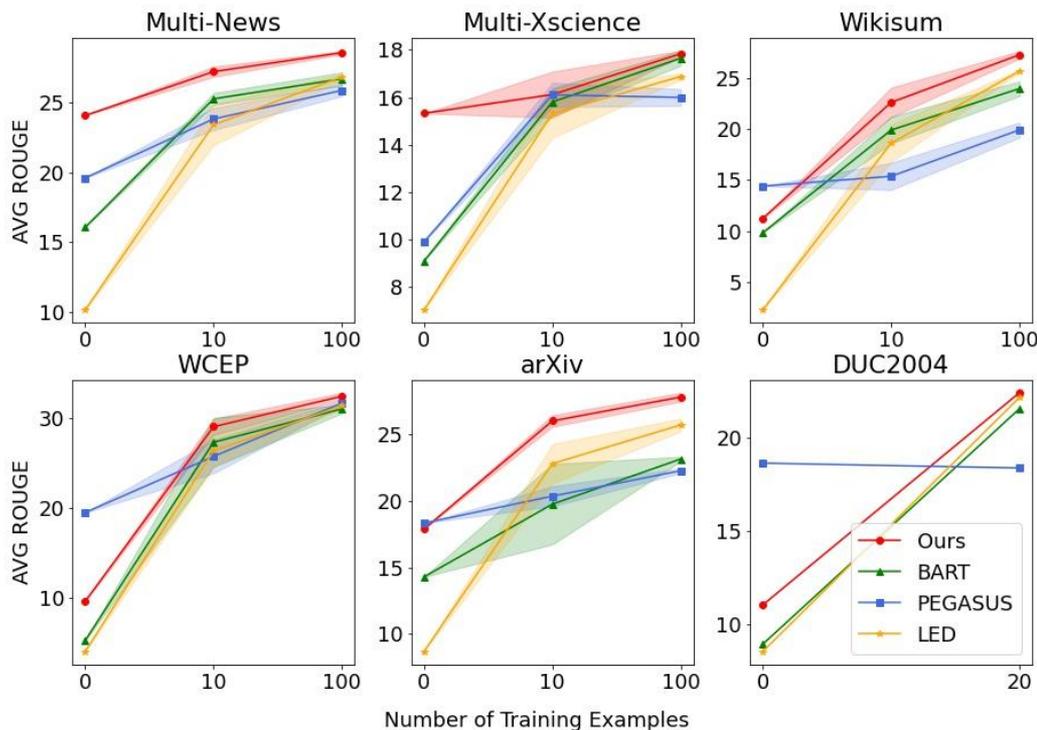
- PEGASUS is also pre-trained for summarization downstream task, thus it performs better than the other two models

Results on Zero-shot



- PEGASUS is also pre-trained for summarization downstream task, thus it performs better than the other two models
- Our model outperforms all the other pre-trained models on most of the datasets (up to 5 ROUGE points)

Results on Few-shot



- Our model outperform all the other pre-trained models on all the datasets.

Results on Fully Supervised

DATASETS	Prev. SOTA			PRIMERA		
	R1	R2	RL	R1	R2	RL
Multi-News	49.2	19.6	24.5	49.9	21.1	25.9
Multi-XScience	34.1	6.8	18.2	31.9	7.4	18.0
WCEP	35.4	15.1	25.6	46.1	25.2	37.9
arXiv	46.6	19.6	41.8	47.6	20.8	42.6

- Our model achieves SOTA on several multi-document summarization datasets, as well a single-document summarization dataset.

Experiments - Human Evaluation

- Datasets:
 - DUC 2007
 - TAC 2008
- Metrics:
 - Pyramid Evaluation
 - Fluency (following DUC guidelines*)

* <https://www-nlpir.nist.gov/projects/duc/duc2007/quality-questions.txt>

Human Evaluation - Pyramid & Fluency

Model	Pyramid Evaluation				Fluency			
	S_r	R	P	F	Gram.	Ref.	Str.&Coh.	
DUC 2007	PEGASUS	6.0	2.5	2.4	2.4	4.45	4.35	1.95
	LED	9.6	3.9	4.0	3.8	4.35	4.50	3.20
	PRIMERA	12.5	5.1	5.0	5.0	4.70	4.65	3.70
TAC 2008	PEGASUS	8.7	9.1	9.4	9.1	4.40	4.20	3.20
	LED	6.9	7.1	10.8	8.4	3.10	3.80	2.55
	PRIMERA	8.5	8.9	10.0	9.3	4.40	4.45	4.10

- PRIMERA also shows a better performance on human evaluation, regarding both pyramid evaluation and fluency evaluation.

- **PRIMERA**, a pre-trained model for **multi-document summarization**.
- It is **pre-trained** with new strategy, **Entity Pyramid**.
- **PRIMERA** reduces the need for dataset-specific architectures and large labeled data.
- **PRIMERA** achieves SOTA on multiple datasets under zero/few-shot and fully supervised, and shows advantage in human evaluation.
- Sample Usage:

```
from transformers import AutoTokenizer, LEDForConditionalGeneration

Tokenizer = AutoTokenizer.from_pretrained('allenai/PRIMERA')
Model = LEDForConditionalGeneration.from_pretrained('allenai/PRIMERA')
```

Code can be found here: <https://github.com/allenai/PRIMER>

Future Work

- Controllable generator to better control the length of generated summaries for zero-shot setting
- Evaluate PRIMERA and its Pyramid Entity strategy on other tasks with multiple documents as input, e.g. Multi-hop QA

Thanks!