# **Final Presentation**

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#### **Problem Statement**

#### Concept Drift

- New Topics / Keywords
- New Summary / Writing Styles
- New Meanings to Words

The **New York Times** in the past

The **New York Times** in the present





It's helpful if a model can be updated, via Incremental Learning

- → First fine-tuned on a large quantity of news summaries
- → Then incrementally updated on small batches of up-to-date news

#### **Problem Statement**

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- New Topics / Keywords
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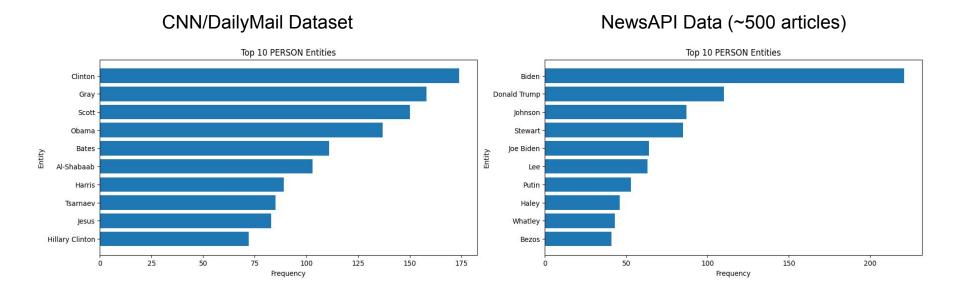
CNN/DailyMail Dataset

#### **Incremental Learning**

- → First fine-tuned on a large quantity of news summaries
- → Then incrementally updated on small batches of up-to-date news

**NewsAPI** 

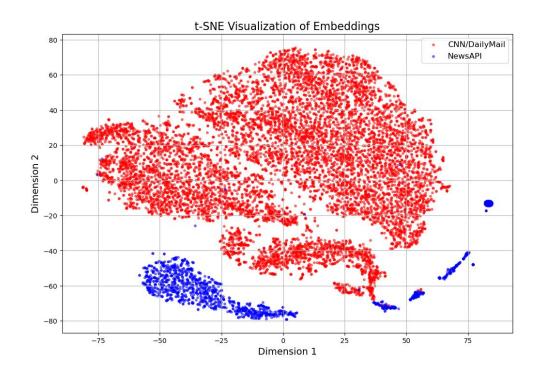
#### **Motivation - Domain Shift Visualized via NER**



# Motivation - Domain Shift Visualized via Embedding

 The last\_hidden\_layer of a bart-large-cnn as the embedding of news articles.

T-SNE clustering



# **Data Exploration - CNN / DailyMail Dataset**

	article	highlights	id	
0	LONDON, England (Reuters) Harry Potter star	Harry Potter star Daniel Radcliffe gets £20M f	42c027e4ff9730fbb3de84c1af0d2c506e41c3e4	11.
1	Editor's note: In our Behind the Scenes series	Mentally ill inmates in Miami are housed on th	ee8871b15c50d0db17b0179a6d2beab35065f1e9	
2	MINNEAPOLIS, Minnesota (CNN) Drivers who we	NEW: "I thought I was going to die," driver sa	06352019a19ae31e527f37f7571c6dd7f0c5da37	
3	WASHINGTON (CNN) Doctors removed five small	Five small polyps found during procedure; "non	24521a2abb2e1f5e34e6824e0f9e56904a2b0e88	
4	(CNN) The National Football League has ind	NEW: NFL chief, Atlanta Falcons owner critical	7fe70cc8b12fab2d0a258fababf7d9c6b5e1262a	

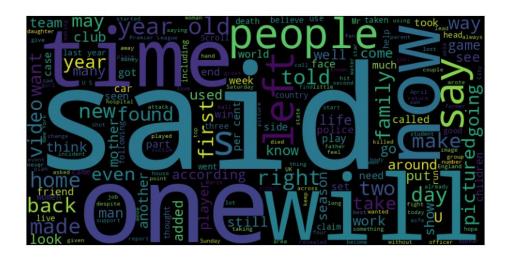
Dataset Split	Number of Instances in Split	
Train	287,113	
Validation	13,368	
Test	11,490	

Feature	Mean Token Count
Article	781
Highlights	56
Total Articles: 287113 Average Article Length: 69	

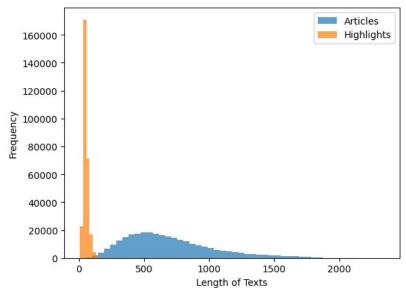
Average Summary Length: 51.574101486174435 words

#### **Visualization**

• Challenges: Colab ran out of RAM trying to visualize 300,000 articles



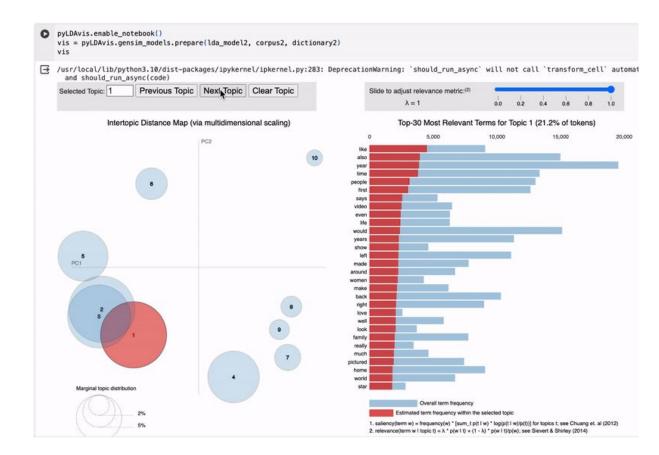
WordCloud frequency for test dataset (11,000 instances)



Article vs Highlights length for whole dataset

# **Topic Modeling**

- Added "said" to stopwords
- Used Latent
   Dirichlet Allocation
   (LDA) model
- Typical topics fairly evident: government/politics, economy/business, entertainment, public health, and sports.



# **NER (Named Entity Recognition) using spaCy**

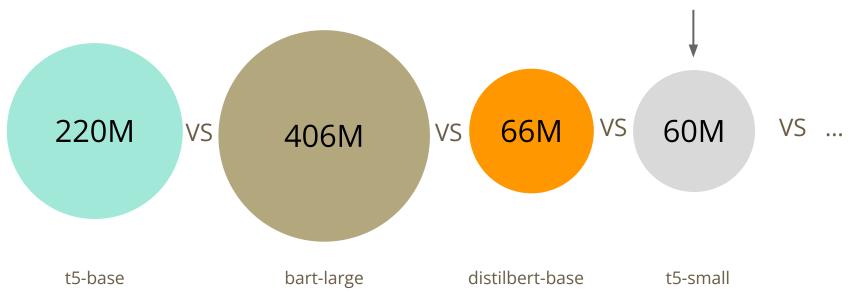
- **PERSON**: People, including fictional.
- NORP: Nationalities, religious or political groups.
- **FAC**: Facilities, like buildings, airports, highways, bridges, etc.
- ORG: Organizations, including companies, agencies, institutions, etc.
- GPE: Geopolitical entities, like countries, cities, states.
- **LOC**: Non-GPE locations, mountain ranges, bodies of water.
- PRODUCT: Objects, vehicles, foods, etc. (Not services.)
- **EVENT**: Named hurricanes, battles, wars, sports events, etc.





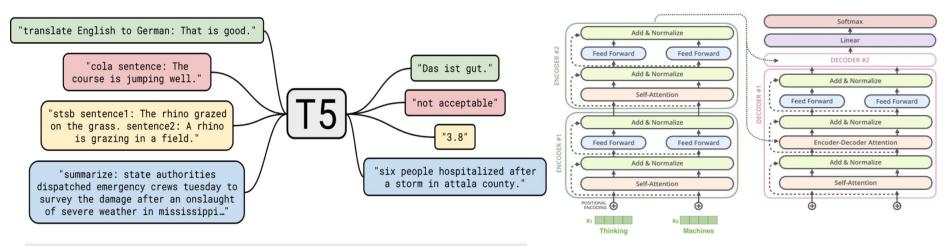


A balance between performance (seq2seq) and memory efficiency (size)



Num of Parameters

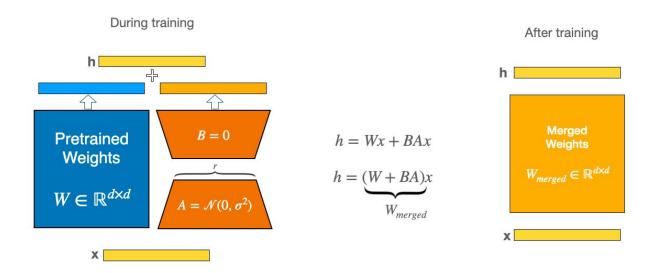
#### **Model Selection**



A diagram of our text-to-text framework. Every task we consider—including translation, question answering, and classification—is cast as feeding our model text as input and training it to generate some target text. This allows us to use the same model, loss function, hyperparameters, etc. across our diverse set of tasks. It also provides a standard testbed for the methods included in our empirical survey. "T5" refers to our model, which we dub the "Text-to-Text Transfer Transformer".

# **LoRA for Model Fine-Tuning**

- Reduce the number of trainable parameters
- Low inference latency
- Model updates are decoupled from the pre-trained weights



#### **Metrics**

#### **Recall-Oriented Understudy for Gisting Evaluation (ROGUE)**

- Compares an automatically produced summary or translation against reference (high-quality and human-produced) summaries or translations.
- Different variants:
  - ROUGE-N: Measures the overlap of n-grams between the system and reference summaries.
  - ROUGE-L: Based on the Longest Common Subsequence (LCS), it accounts for sentence-level structure similarity.

#### **Metrics**

$$F1 = \frac{2 \times Precision \times Recall}{(Precision + Recall)}$$

**Precision**: # n-grams in S that appear also in A / # n-grams in S

Recall: # n-grams in A that appear also in S / # n-grams in A

<b>Article (A):</b> The cat is on the mat.	<b>Summary (S):</b> The cat and the dog.
--	--

**ROUGE-1**: Precision = 3/5 = 0.6 Recall = 3/6 = 0.5 F1-score = 0.55

**ROUGE-2**: Precision = 1/4 = 0.25 Recall = 1/5 = 0.20 F1-score = 0.22

**ROUGE-L**: Precision = 3/5 = 0.6 Recall = 3/6 = 0.5 F1-score = 0.55

## **News Retrieval Pipeline**

- Fully automated and hosted on Google Cloud.
- Scheduled to automatically query NewsAPI, a service that gathers the latest news from numerous sources (NYT, WSJ, etc).



## **News Retrieval Pipeline**

- Processes the latest news articles, on average numbering ~500 articles / day.
  - NewsAPI provides a "description" for each article, which is used as the summary
  - The article url is scraped for its content.
  - Then, the summary and content is grouped as data and fed into our model for the incremental learning to take into effect.

```
Top business headlines in the US right now
       https://newsapi.org/v2/top-headlines?country=us&category=business&ap
       iKev=API KEY
 status: "ok",
 totalResults: 67,
- articles: [
      - source:
        author: "Ramishah Maruf, Pete Muntean, Gregory Wallace",
        title: "Alaska Airlines CEO says company found loose bolts on 'many'
        description: "Alaska Airlines CEO Ben Minicucci revealed the carrier found
        "some loose bolts on many" Boeing 737 Max 9s in an interview for "NBC
       Nightly News with Lester Holt" scheduled to air Tuesday.",
       url: https://www.cnn.com/2024/01/23/business/alaska-airlines-ceo-boeing-ma
        x-9-loose-bolts/index.html.
        urlToImage: https://media.cnn.com/api/v1/images/stellar/prod/240123161554
        ben-minicucci-ceo-alaska-airlines-09152022.jpg?c=16x9&g=w 800,c fill,
        publishedAt: "2024-01-24T00:22:30Z".
        content: "Alaska Airlines CEO Ben Minicucci revealed the carrier found
        some loose bolts on many Boeing 737 Max 9s in an interview for NBC Nightly
        News with Lester Holt scheduled to airTuesday.\r\nIts the CEOs fi... [+3538
        chars]"
      - source:
           name: "Yahoo Entertainment"
```

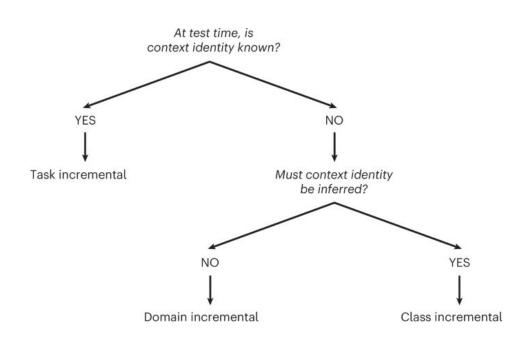
# **News Retrieval Pipeline**

- The news sources are carefully selected and filtered
  - Considering the different biases and reputation
- The pipeline also has built-in error handling
  - Able to continue processing data even if faulty links or data is encountered.

# Literature Review on Incremental Learning

Three types of Incremental Learning:

- 1. Task Incremental
- 2. Domain Incremental (Our task)
- 3. Class Incremental



# **But, IL/CL Comes with Catastrophic Forgetting**

#### **Catastrophic Forgetting**

Models might perform worse on older datasets as they are more and more fine-tuned on new datasets.

#### Solutions Attempted

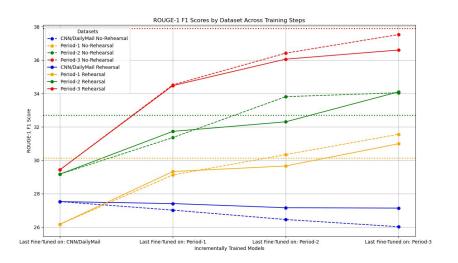
**Experimental Rehearsal / Replay:** storing previous samples and reusing them when training on new samples.

Elastic Weight Consolidation (EWC): While learning a new task, EWC protects the performance on an old task by constraining the parameters to stay in a region of low error for the old task.

# **Experiments with Rehearsal/Replay**

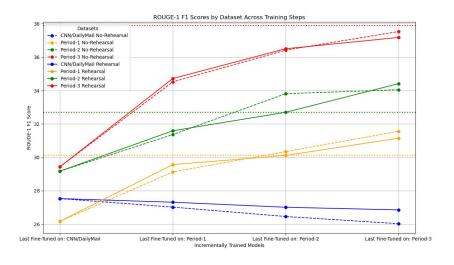
Rehearsal vs. No Rehearsal

(33% old data)



#### Rehearsal vs. No Rehearsal

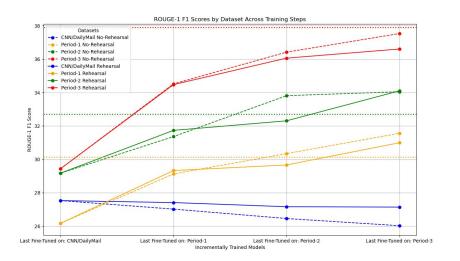
(16% old data)



# **Experiments with Rehearsal/Replay**

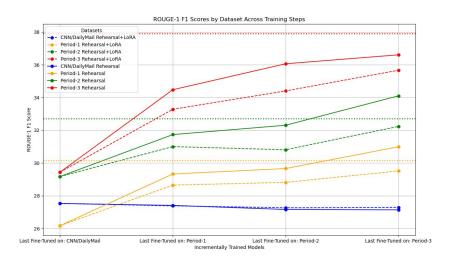
Rehearsal vs. No Rehearsal

(33% old data)



#### Rehearsal vs. LoRA Rehearsal

1764 mb vs. 938 mb

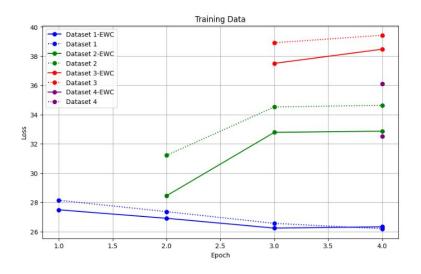


# **Experiments with EWC**

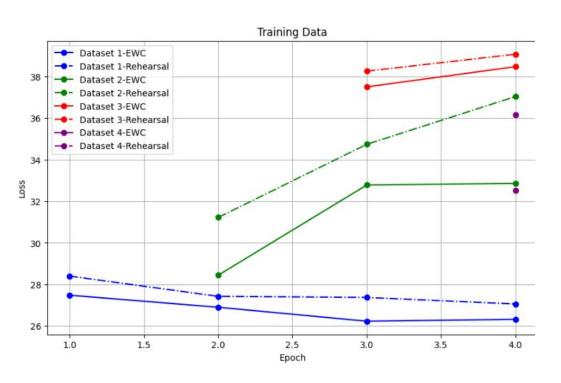
#### **EWC Quadratic Penalty**

$$\mathcal{L}(\theta) = \mathcal{L}_B(\theta) + \sum_i \frac{\lambda}{2} F_i (\theta_i - \theta_{A,i}^*)^2$$
(3)

where  $\mathcal{L}_{\mathcal{B}}(\theta)$  is the loss for task B only,  $\lambda$  sets how important the old task is compared to the new one and i labels each parameter.



# **Experiments with EWC**



#### Rehearsal vs EWC

 Regularization (EWC) doesn't perform as well as Rehearsal method, as we expected before the experiment based on related works.

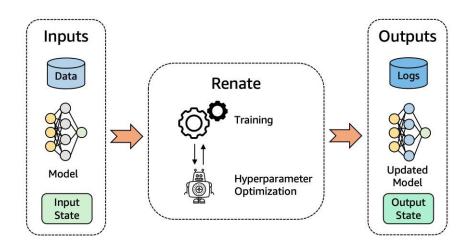
 Adopt the REHEARSAL strategy for continual learning for the final deployed model

# **Continual Learning API**

# Avalanche: an End-to-End Library for Continual Learning

Powered by ContinualAI





#### **Trial on Avalanche and Renate**

- Find compatible CL packages for Huggingface and PyTorch
- Run some demos of CL at Colab

```
from typing import Callable, Dict, Optional
 import torch
 from torchmetrics import Accuracy
 from torchvision.transforms import transforms
 from renate import defaults
 from renate.benchmark.datasets.vision_datasets import TorchVisionDataModule
 from renate.benchmark.models.mlp import MultiLayerPerceptron
 from renate.benchmark.scenarios import ClassIncrementalScenario. Scenario
 from renate.models import RenateModule
 def data_module_fn(data_path: str, chunk_id: int, seed: int = defaults.SEED) -> Scenario:
     """Returns a class-incremental scenario instance.
     The transformations passed to prepare the input data are required to convert the data to
     PvTorch tensors.
     data_module = TorchVisionDataModule(
         data_path,
        dataset_name="MNIST",
        val_size=0.1,
         seed=seed.
     class incremental scenario = ClassIncrementalScenario(
        data module=data module,
        groupings=((0, 1, 2, 3, 4), (5, 6, 7, 8, 9)),
        chunk_id=chunk_id,
     return class incremental scenario
 def model fn(model state url: Optional[str] = None) -> RenateModule:
```

```
Avg acc: 37.743333333333333
Training on task: 1
Train Epoch: 1 Loss: 1.987624
Train Epoch: 2 Loss: 1.593181
Testing on task: 0
Test set: Average loss: 0.0062, Accuracy: 5666/10000 (57%)
Testing on task: 1
Test set: Average loss: 0.0034, Accuracy: 7701/10000 (77%)
Testing on task: 2
Test set: Average loss: 0.0114, Accuracy: 1164/10000 (12%)
Avg acc: 48.4366666666667
Training on task: 2
Train Epoch: 1 Loss: 2.034442
Train Epoch: 2 Loss: 1.997389
Testing on task: 0
Test set: Average loss: 0.0082, Accuracy: 5225/10000 (52%)
Testing on task: 1
Test set: Average loss: 0.0068, Accuracy: 4019/10000 (40%)
Testing on task: 2
Test set: Average loss: 0.0042, Accuracy: 6989/10000 (70%)
Avg acc: 54.10999999999999
```

#### **Trial on Avalanche and Renate**

#### An unfortunate result:

- Conflicts with the version of multiple python packages
- Did not meet our expectations (improvement, workload)
- Few references online (haven't updated since Jan 2023)

# **HuggingFace Space**

Online and running:

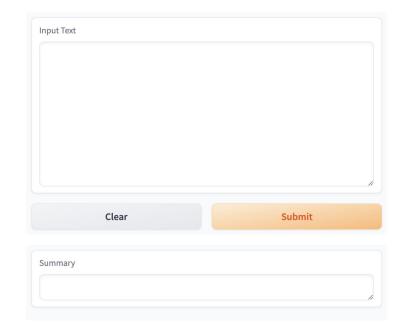
https://huggingface.co/spaces/liamvbetts/b art-news-summary-v1

```
    app.py 2 X ≡ requirements.txt

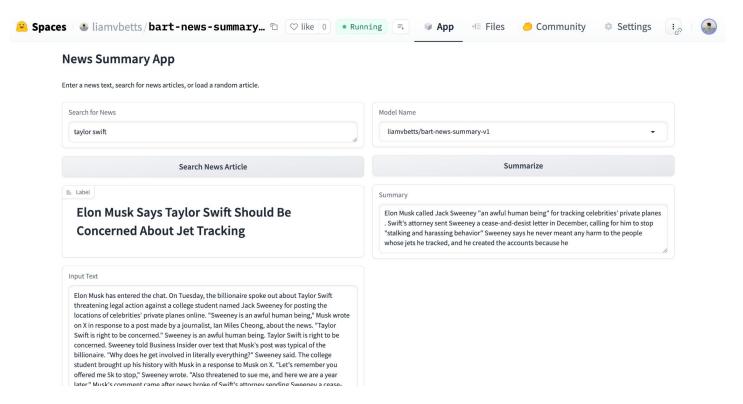
♦ app.py > ...
       import gradio as gr
       from transformers import AutoTokenizer, AutoModelForSeq2SeqLM
       tokenizer = AutoTokenizer.from_pretrained("liamvbetts/bart-large-cnn-v4")
       model = AutoModelForSeq2SeqLM.from_pretrained("liamvbetts/bart-large-cnn-v4")
      def summarize(article):
           inputs = tokenizer(article, return_tensors="pt").input_ids
           outputs = model.generate(inputs, max_new_tokens=128, do_sample=False)
           summary = tokenizer.decode(outputs[0], skip_special_tokens=True)
           return summary
       input text = gr.Textbox(lines=10, label="Input Text")
       output_text = gr.Textbox(label="Summary")
      gr.Interface(
          inputs=input_text,
          outputs=output text,
           title="News Summary App",
           description="Enter a news text and get its summary."
      ).launch()
```



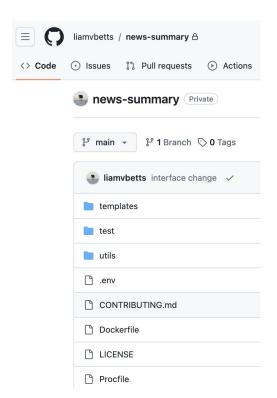
Enter a news text and get its summary.

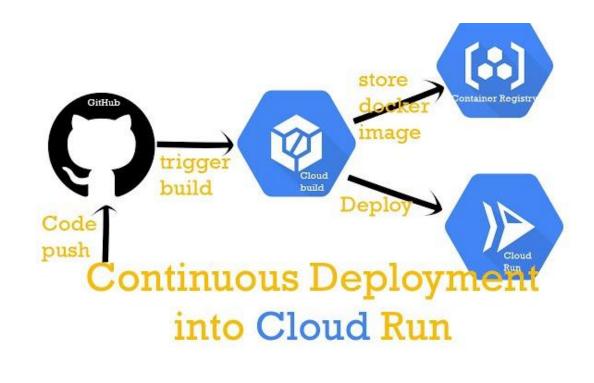


# **HuggingFace Space**



# **Cloud Run with Continuous Deployment**

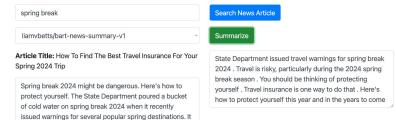




## Flask App

- Updated / organized UI
- Button to fetch today's fine-tuned model
- Now calculates ROUGE scores between NewsAPI highlights and generated summaries

#### **News Summary App**

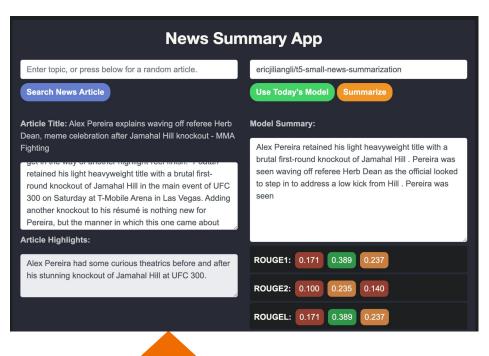


#### Results:

Word Count - Article: 512

also got everyone to start thinking about travel insurance,

Word Count - Summary: 52



# **Daily Automated Script**

- Integrated data retrieval from NewsAPI with retraining code
- When run, this code:
  - Pulls 500 articles from NewsAPI into a pandas dataframe
  - Pulls the latest model on HuggingFace
  - Fine tunes it (has LoRA capability)
  - Pushes it to HF
  - Deletes it locally
- Using Colab Pro+ Scheduling

```
1 Micapture
2 | pip install transformers(torch) accelerate —U
3 | pip install datasety|
4 | pip install roupe_score
5 | pip install pert
6 | pip install newspaper3k
7 | pip install newspaper3k
       1 from transformers import AutoTokenizer, AutoModelForSeq2Seq1M, DataCollatorForSeq2Seq, Seq2SeqTrainingArguments, Seq2SeqTraine
             import pandas as pd
from sklearn.model_selection import train_test_split
               sport requests
inport ReadTimeout, ConnectTimeout, HTTPError, Timeout, ConnectionError
inp peft import (
                  prepare model for kbit training.
             newsap1 = NewsApiClient(ap1_key='53f9ca03b7c14c858c2f072fe04b180c'
1   data = []
2   today_date = datetime.today().date()
3   yesterday_date = today_date - timedetta(days=3)
            cur date = vesterday date.strftime("ay-am-ad")
                            ad = {
   'content': fulltext(r.text),
                              r ad['summary'] == 'NaN':
                           data, append(ad)
                        except (ConnectTimeout, MTTPError, ReadTimeout, Timeout, ConnectionError):
                                                                     from parametur date
                                                                    to=cur_date,
language='en',
sert_by='popularity',
page=1)
                  add_new_articles(all_articles('articles'))
                  page = 1
while page*100 < tr:
                       all articles = newsapi.get everything(
                                                                    from_parametur_date,
to=cur_date,
language='en',
sort_byw'popularity',
                       add new articles(all articles('articles'))
                  if not os.path.exists('./datasets'):
                  df.to_csv(csv_name, index=False)
```

```
# configuration
metric = loae_metric("ro
prefix = "summarize) "
mx_laput_length = 512
mx_larget_length = 64
       riletame = get_mewsapi_data()
if filetame is fame;
prant("metraceong aported due to lack of data."
       # Check If detainant is empty after loading
if of empty:
print/"hetrizing sourted the Dutainant is empty after landing data.")
roturn News
       df = df.eropna(subset=('summary'))
df = df(ef('content').apply(lambda x; len(x) >= 500))
              USEK IT down as copy of the format of the format of the cleaning and filtering.")
print("Metraining aborted: The DataFrame is empty after cleaning and filtering.")
```

### A Showcase of IL/CL's Effectiveness

Rouge scoring averaged over ~500 news articles released in 2024

	T5 + Incremental Learning	T5
Rouge-1	0.151605	0.098356
Rouge-2	0.110561	0.077415
Rouge-I	0.142509	0.093274

## A Case Study on IL/CL's Effectiveness

**Article:** Five crew members onboard the submersible Titan were probably killed instantly in a "catastrophic implosion" as it descended to the wreck of the Titanic two miles below the surface of the Atlantic ocean, US Coast Guard officials announced on Thursday. A large debris field containing multiple sections of the vessel was spotted earlier in the day by a remotely operated vehicle (ROV) scouring the seabed near the Titanic wreck site 400 miles south of St John's, Newfoundland, officials said at an afternoon press conference in Boston

T5 + IR: US Coast Guard officials announced Thursday five crew members onboard the submersible Titan were probably killed instantly in a "catastrophic implosion" as it descended to the wreck of the Titanic two miles below the surface of the Atlantic Ocean. A large debris field containing multiple sections of the vessel was spotted earlier in the day by a remotely operated vehicle (ROV) (0.686046511627907, 0.6588235294117647, 0.6046511627906976)

T5: US Coast Guard officials announced on Thursday that five crew members onboard the submersible Titan were killed instantly in a "catastrophic implosion" as it descended to the wreck of the Titanic two miles below the surface of the Atlantic ocean. Five crew members onboard the submersible Titan were probably killed instantly in a "catastrophic implosion" as it descended to the wreckage, US Coast Guard officials said. Five crew members onboard the submersible Titan were probably killed instantly in afloat.

(0.5232558139534884, 0.4235294117647059, 0.5116279069767442)









#### **Division of Labor**

- **Eric**: Experiments with rehearsal, LoRA & hyperparameter fine-tuning, domain shift exploration & visualization
- Jack: Experiments with rehearsal & EWC, literature review on IL/CL, Huggingface framework integration
- Kyle: NewsAPI integration, automated data pipeline implementation, data preprocessing & filtering
- **Liam**: Data exploration & visualization, model fine tuning, HuggingFace deployment, Flask app, automated data pipeline implementation
- Kun: Literature Review on IL/CL, model selection, Avalanche & Renate implementation

# Thank you!