# A QUESTION-ORIENTED PROPAGATION NETWORK FOR NEWS READING COMPREHENSION Yingwei Luo<sup>1,2</sup>

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# Overview

**Task:** Build an question answering model that can read a news article and answer questions related to it.

# **Example:**

**Question:** Where is Brittanee Drexel from?

**Article:** The mother of a 17-year-old Rochester, New York, high school student who vanished over the weekend on spring break in Myrtle Beach, South Carolina, says she did not give her daughter permission to go on the trip... Brittanee Marie Drexel's mom says she thought she was at the beach in New York, not South...

**Answer:** Rochester, New York

# **Challenges:**

- News articles usually are long while the maximum input length of state-of-the-art question answering (QA) models such as BERT and RoBERTa is limited to 512.
- To answer a question, one need to synthesize information across different parts of an article.

# **Previous Approaches:**

1) Sliding Window Technique: Limited by window size.

2) Coarse-to-Fine Paradigm: Not suitable for QA tasks that contain long answers that span multiple sentences.

3) Sparse Attention Mechanism: Rely on pre-defined hand-designed attention patterns.

# **Datasets:**

- \* NewsQA (Trischler et al., 2017)
- \* NLQuAD (Soleimani et al., 2021)
- \* Quasar-T (Dhingra et al., 2017)

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### Method

# **Model Components:**

- Context Encoding Module
- Multi-step Reasoning Module
  - Question-Oriented Information Interaction
  - 2. Gate-Based Information Fusion
  - 3. Question-Guided Information Propagation
- Answer Prediction Module

# **Model Architecture:**



# **Advantages:**

- 1. Take the whole article into consideration and jointly learn to find question-related clues and make inference over them implicitly.
- 2. Does not rely on hand-designed patterns and directly aims at question-focused information.



# Results

#### **Performance on NewsQ**

Model FastQAExt AMANDA MINIMAL DECAPROP **RoBERTa-large** (sli CogLTX **QOPN** (RoBERTa-QOPN

### Performance on NLQuA

Model BERT-base **BERT-large** RoBERTa-base **RoBERTa-large** Longformer QOPN

# Ablations

Model

- QOPN(full model)
- Question-oriented
- Gate-based fusior
- Question-guided

# Effect of Number of Rea







A (Trischler et al., 2017)				
	EM(%	%) F1	(%)	
	42.8	3 5	6.1	
	48.4	l 6	3.7	
	50.1	. 6	3.2	
	53.1	. 6	6.3	
liding window)	49.6	6 6	6.3	
	55.2	2 7	0.1	
-base)	61.2	2 7	5.1	
	65.5	5 7	9.8	
D (Soleimani et al., 2021)				
EM(%	b) F1(	%) la	oU(%)	
25.0	64	.0	53.8	
30.3	67	.9	58.4	
29.1	67	.2	57.7	
33.4	71	.1	62.4	
50.3	81	.4	73.6	_
<b>54.0</b>	82	.9	<b>75.8</b>	_
	EM	F1	loU	
	54.0	82.9	<b>75.8</b>	
l interaction	47.0	79.4	71.2	
1	52.8	82.3	75.1	
oropagation	51.2	80.9	73.3	
asoning Blocks:				
82.9	91 /			
79.8	<u>81</u> .4			
78.5	78.8			