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Retrieve, Rerank and Rewrite: Soft Template Based Neural Summarization

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Sentence Summarization

Definition

- Generate a shorter version of a given sentence
- Preserve its original meaning

Usage

Design or refine appealing headlines

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Seq2seq Sum	marization		

- Require less human efforts
- Achieve the state-of-the-art performance



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Problems of Seq2	seq Summarizatio	on	

Solely depend on the source text to generate summaries

- Encounter error propagation
- Lose control
 - 3% of summaries \leq 3 words
 - 4 summaries repeat a word for 99 times
 - Focus on extraction rather than abstraction

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Template-based S	ummarization		

- A traditional approach to abstractive summarization
- Fill an incomplete with the input text using the manually defined rules
- Be able to produce fluent and informative summaries

Template	[REGION] shares [open/close] [NUMBER]		
	percent [lower/higher]		
Source	hong kong shares closed down $#.#$ percent on		
	friday due to an absence of buyers and fresh		
	incentives .		
Summary	<i>hong kong</i> shares <i>close</i> #.# percent <i>lower</i>		

Problems of	Template-based	Summarization	
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- Template construction is extremely time-consuming and requires a plenty of domain knowledge
- It is impossible to develop all templates for summaries in various domains

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Motivation			

Use actual summaries in the training datasets as **soft templates** to combine seq2seq and template-based summarization Seq2seq Guide the generation of seq2seq Template-based Automatically learn to rewrite from soft templates

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Proposed Method	d		

Re³Sum: consists of three modules: **Re**trieve, **Re**rank and **Re**write.

- Use Information Retrieval to find out candidate soft templates from the training dataset (Retrieve).
- Extend the seq2seq model to jointly learn template saliency measurement (Rerank) and final summary generation (Rewrite)

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Contributions			

- Introduce soft templates to improve the readability and stability in seq2seq
- Extend seq2seq to conduct template reranking and template-aware summary generation simultaneously
- Fuse the IR-based ranking technique and seq2seq-based generation technique, utilizing both supervisions
- Oemonstrate potential to generate diversely

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Flow Chat			

- Retrieve Search actual summaries as candidate soft templates
 - Rerank Find out the most proper soft template from the candidates
- Rewrite Generate the summary based on source sentence and soft template



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Retrieve			

Assumption: Similar sentences, similar summary patterns

Input A sentence

Platform LUCENE

Output 30 actual summaries in the training dataset whose sources are the most similar to the input sentence

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Jointly Rera	nk and Rewrite		

Share encoders



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Rerank			

- Retrieve ranks templates according to the text similarity between sentences
- Rerank finds out the soft template most similar to the actual output summary

Model: Bilinear network

$$s(\mathbf{r}, \mathbf{x}) = \text{sigmoid}(\mathbf{h}_r \mathbf{W}_s \mathbf{h}_x^{\mathsf{T}} + b_s)$$

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Rewrite			

- A soft template accords with the facts in the input sentences
- Use Seq2seq to generate more faithful and informative summaries

Concatenate the encoders of sentence and template

$$\mathbf{H}_{c} = [\mathbf{h}_{1}^{x}; \cdots; \mathbf{h}_{-1}^{x}; \mathbf{h}_{1}^{r}; \cdots; \mathbf{h}_{-1}^{r}]$$

Use attentive RNN decoder to generate summaries

$$\mathbf{s}_t = \mathsf{Att}\mathsf{-}\mathsf{RNN}(\mathbf{s}_{t-1}, y_{t-1}, \mathbf{H}_c),$$

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Learning			

- Cross Entropy (CE) for Rerank
- Negative Log-Likelihood (NLL) for Rewrite
- Add the above two costs as the final loss

$$J_{R}(\theta) = \operatorname{CE}(s(\mathbf{r}, \mathbf{x}), s^{*}(\mathbf{r}, \mathbf{y}^{*}))$$

$$= -s^{*} \log s - (1 - s^{*}) \log(1 - s)$$

$$J_{G}(\theta) = -\log(p(\mathbf{y}^{*}|\mathbf{x}, \mathbf{r}))$$

$$= -\sum_{t} \log(\mathbf{p}_{t}[y_{t}^{*}])$$

$$J(\theta) = J_{R}(\theta) + J_{G}(\theta)$$

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Setting			

Dataset Gigaword (sentence, headline) pairs Framework OpenNMT

Dataset	Train	Dev.	Test
Count	3.8M	189k	1951
AvgSourceLen	31.4	31.7	29.7
AvgTargetLen COPY(%)	8.3	8.3	8.8
COPY(%)	45	46	36

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ROUGE Performa	ance		

Re³Sum significantly outperforms other approaches

Model	ROUGE-1	ROUGE-2	ROUGE-L
ABS [†]	29.55*	11.32*	26.42*
$ABS+^\dagger$	29.78*	11.89*	26.97*
$Featseq2seq^\dagger$	32.67*	15.59*	30.64*
$RAS ext{-}Elman^\dagger$	33.78*	15.97*	31.15*
$Luong-NMT^{\dagger}$	33.10*	14.45*	30.71*
OpenNMT	35.01*	16.55*	32.42*
Re ³ Sum	37.04	19.03	34.46

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Linguistic Quality	Performance		

- $\bullet~\mbox{Lense}$ Low LEN_DIF and LESS_3 $\rightarrow~\mbox{Stable}$
- Low COPY \rightarrow Abstractive
- $\bullet~\mbox{Low NEW_NE}$ and $\mbox{NEW_UP} \to \mbox{Faithful}$

ltem	Template	OpenNMT	Re ³ Sum
LEN_DIF	2.6±2.6	3.0±4.4	2.7±2.6
LESS_3	0	53	1
COPY(%)	31	80	74
NEW_NE	0.51	0.34	0.30
NEW_UP	0.38	0.19	0.11

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Effects of Template				

- Performance highly relies on templates
- The rewriting ability is strong

Туре	ROUGE-1	ROUGE-2	ROUGE-L
+Random	32.60	14.31	30.19
+First	36.01	17.06	33.21
+Max	41.50	21.97	38.80
+Optimal	46.21	26.71	43.19
$+ Rerank(Re^3Sum)$	37.04	19.03	34.46

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Generation Diversity				

OpenNMT Beam search n-best outputs Re³Sum Provide different templates

Source	anny ainge said thursday he had two one-hour		
	meetings with the new owners of the boston celtics		
	but no deal has been completed for him to return		
	to the franchise .		
Target	ainge says no deal completed with celtics		
Templates	major says no deal with spain on gibraltar		
Templates	roush racing completes deal with red sox owner		
Re ³ Sum	ainge says no deal done with celtics		
Re Sum	ainge talks with new owners		
OpenNMT	ainge talks with celtics owners		
	ainge talks with new owners		

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Conclusion			

- Introduce soft templates as additional input to guide seq2seq summarization
- Combine IR-based ranking techniques and seq2seq-based generation techniques to utilize both supervisions
- Improve informativeness, stability, readability and diversity

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Thank you