

Pretraining-Based Natural Language Generation for Text Summarization

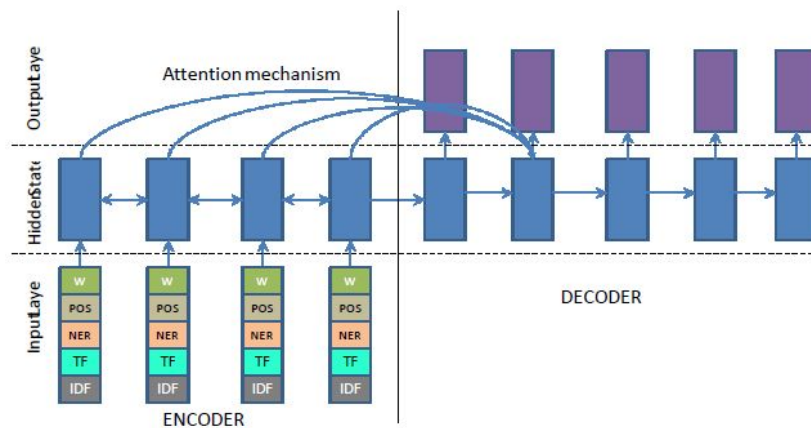
Yixian Liu

2019/4/17

Text summarization

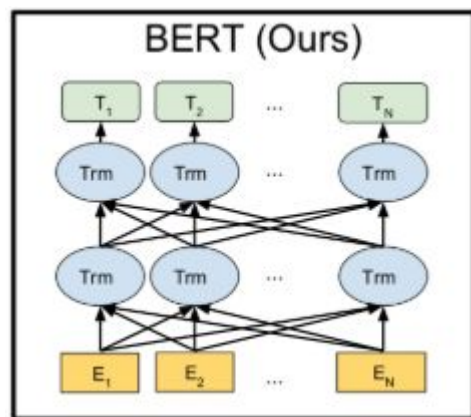
seq-to-seq model

from a document to its summarization

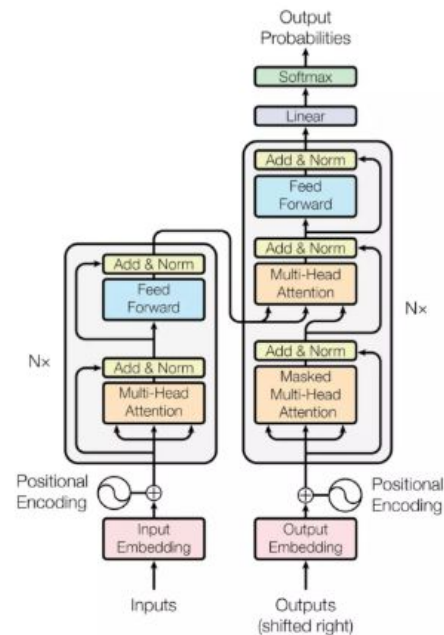


Bert and transformer

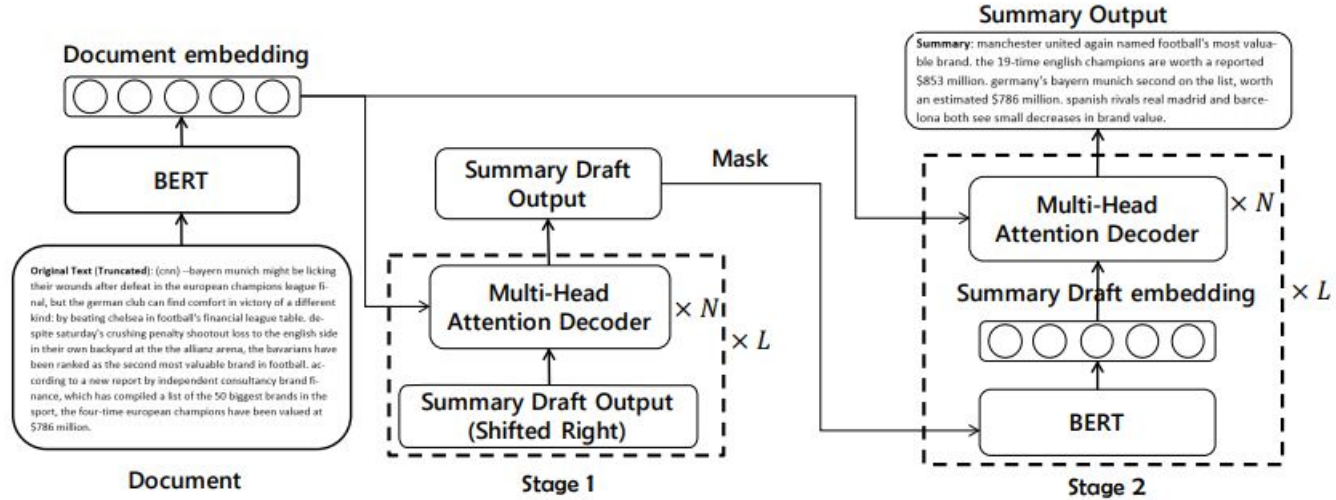
Bert



Transformer



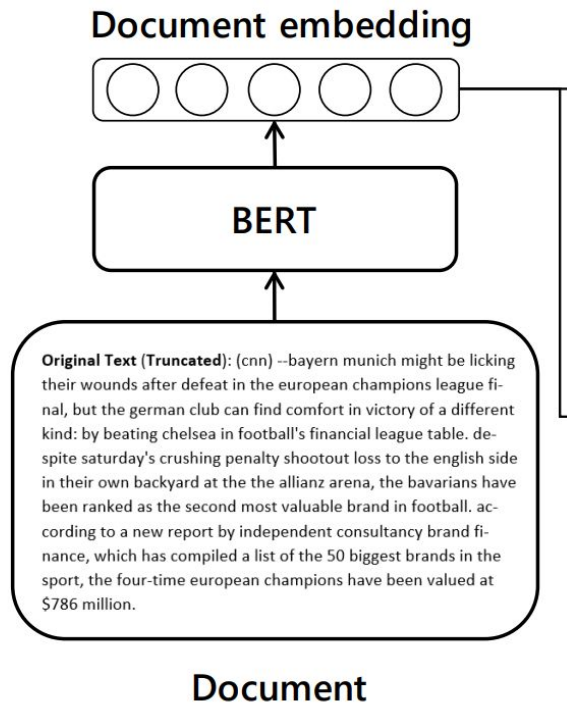
Pretraining-Based model



Encoder

$$H = BERT(x_1, \dots, x_m)$$

$$\hat{H} = \{h_1, \dots, h_m\}$$



Stage I decoder

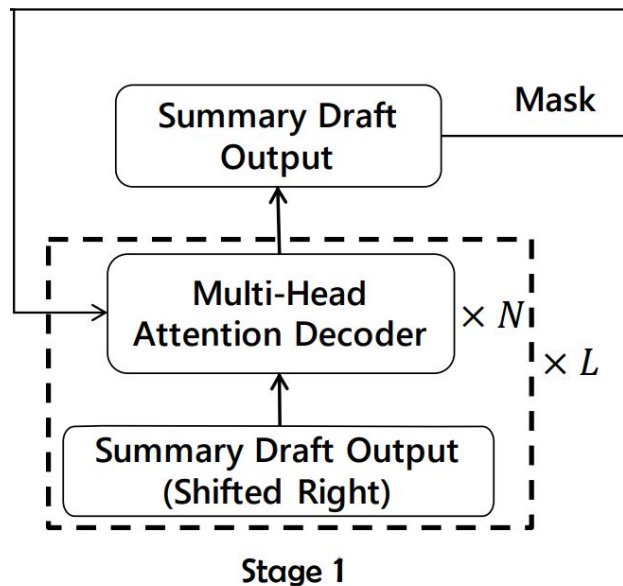
Time step t

$$P_t^{vocab}(w) = f_{dec}(q_{<t}, H)$$

$$L_{dec} = \sum_{t=1}^{|a|} -\log P(a_t = y_t^* | a_{<t}, H)$$

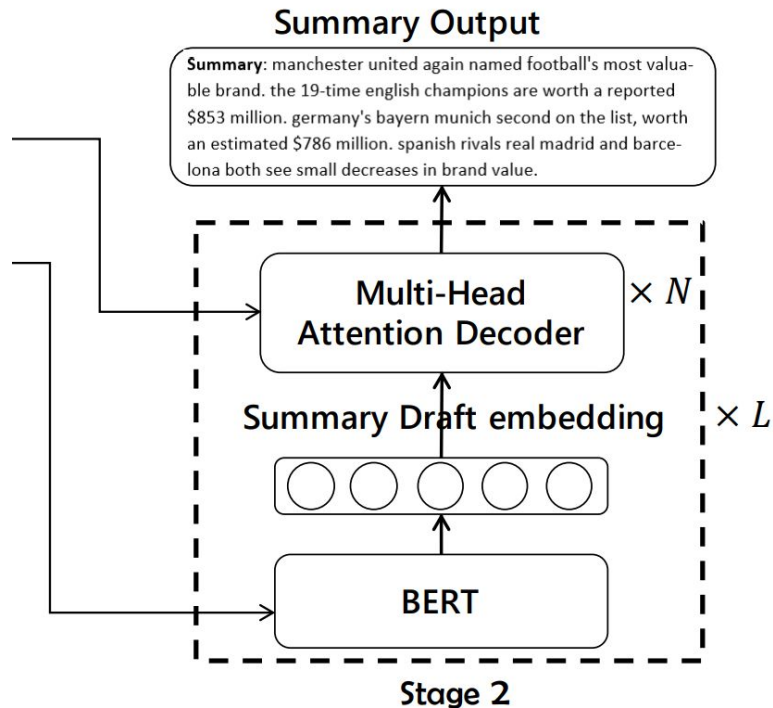
Copy mechanism

$$P_t(w) = (1 - g_t)P_t^{vocab}(w) + g_t \sum_{i:w_i=w} \alpha_t^i$$



Stage II decoder ---- Summary Refine

$$L_{refine} = \sum_{t=1}^{|y|} -\log P(y_t = y_t^* | a_{\neq t}, H)$$



Mixed objective

Policy gradient about ROUGE

$$L_{dec}^{rl} = R(a^s) \cdot [-\log(P(a^s|x))]$$
$$\hat{L}_{dec} = \gamma * L_{dec}^{rl} + (1 - \gamma) * L_{dec}$$

Final objective function

$$L_{model} = \hat{L}_{dec} + \hat{L}_{refine}$$

Result

Model	ROUGE-1	ROUGE-2	ROUGE-L	R-AVG
Extractive				
lead-3 [See <i>et al.</i> , 2017]	40.34	17.70	36.57	31.54
SummmaRuNNer [Nallapati <i>et al.</i> , 2017]	39.60	16.20	35.30	30.37
Refresh [Narayan <i>et al.</i> , 2018]	40.00	18.20	36.60	31.60
DeepChannel [Shi <i>et al.</i> , 2018]	41.50	17.77	37.62	32.30
rnn-ext + RL [Chen and Bansal, 2018]	41.47	18.72	37.76	32.65
MASK- LM^{global} [Chang <i>et al.</i> , 2019]	41.60	19.10	37.60	32.77
NeuSUM [Zhou <i>et al.</i> , 2018]	41.59	19.01	37.98	32.86
Abstractive				
PointerGenerator+Coverage [See <i>et al.</i> , 2017]	39.53	17.28	36.38	31.06
ML+RL+intra-attn [Paulus <i>et al.</i> , 2018]	39.87	15.82	36.90	30.87
inconsistency loss[Hsu <i>et al.</i> , 2018]	40.68	17.97	37.13	31.93
Bottom-Up Summarization [Gehrmann <i>et al.</i> , 2018]	41.22	18.68	38.34	32.75
DCA [Celikyilmaz <i>et al.</i> , 2018]	41.69	19.47	37.92	33.11
Ours				
One-Stage	39.50	17.87	36.65	31.34
Two-Stage	41.38	19.34	38.37	33.03
Two-Stage + RL	41.71	19.49	38.79	33.33

Table 1: ROUGE F1 results for various models and ablations on the CNN/Daily Mail test set. R-AVG calculates average score of Rouge-1, Rouge-2 and Rouge-L.

Result

Model	R-1	R-2
First sentences	28.6	17.3
First k words	35.7	21.6
Full [Durrett <i>et al.</i> , 2016]	42.2	24.9
ML+RL+intra-attn [Paulus <i>et al.</i> , 2018]	42.94	26.02
Two-Stage + RL (Ours)	45.33	26.53

Table 2: Limited length ROUGE recall results on the NYT50 test set.