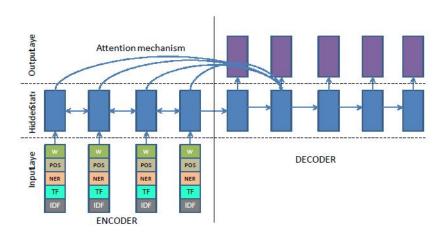
# Pretraining-Based Natural Language Generation for Text Summarization

Yixian Liu 2019/4/17

### Text summarization

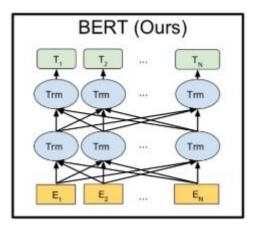
seq-to-seq model

from a document to its summarization

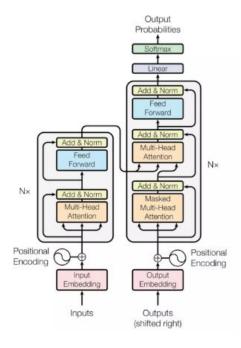


#### Bert and transformer

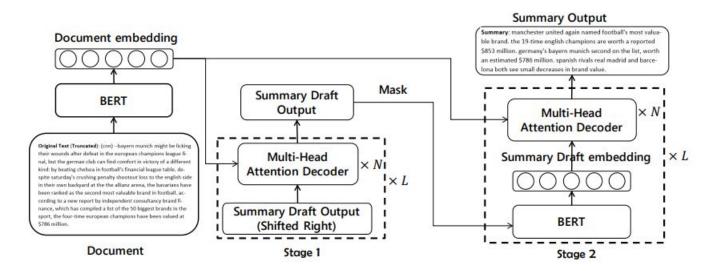
Bert



Transformer



# **Pretraining-Based model**

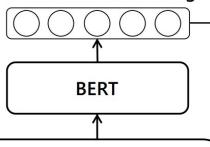


#### Encoder

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

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

#### **Document embedding**



Original Text (Truncated): (cnn) --bayern munich might be licking their wounds after defeat in the european champions league final, but the german club can find comfort in victory of a different kind: by beating chelsea in football's financial league table. despite saturday's crushing penalty shootout loss to the english side in their own backyard at the the allianz arena, the bavarians have been ranked as the second most valuable brand in football. according to a new report by independent consultancy brand finance, which has compiled a list of the 50 biggest brands in the sport, the four-time european champions have been valued at \$786 million.

#### **Document**

## Stage I decoder

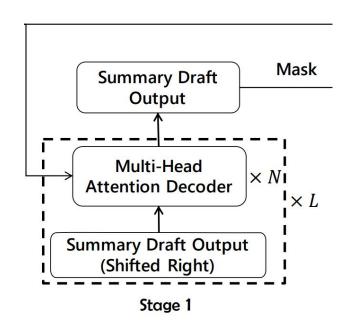
Time step t

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

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

Copy mechanism

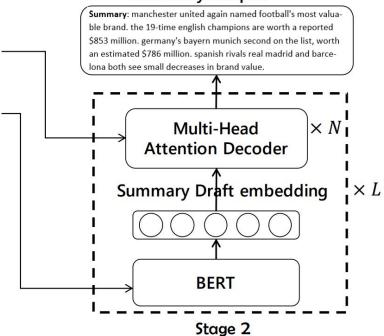
$$P_t(w) = (1 - g_t)P_t^{vocab}(w) + g_t \sum_{i:w_i = w} \alpha_t^i$$



# Stage II decoder ---- Summary Refine

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

#### **Summary Output**



# Mixed objective

Policy gradient about ROUGE

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

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 et al., 2017]                       | 40.34   | 17.70   | 36.57   | 31.54 |
| SummmaRuNNer [Nallapati et al., 2017]           | 39.60   | 16.20   | 35.30   | 30.37 |
| Refresh [Narayan et al., 2018]                  | 40.00   | 18.20   | 36.60   | 31.60 |
| DeepChannel [Shi et al., 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 et al., 2019]        | 41.60   | 19.10   | 37.60   | 32.77 |
| NeuSUM [Zhou et al., 2018]                      | 41.59   | 19.01   | 37.98   | 32.86 |
| Abstractive                                     |         |         |         |       |
| PointerGenerator+Coverage [See et al., 2017]    | 39.53   | 17.28   | 36.38   | 31.06 |
| ML+RL+intra-attn [Paulus et al., 2018]          | 39.87   | 15.82   | 36.90   | 30.87 |
| inconsistency loss[Hsu et al., 2018]            | 40.68   | 17.97   | 37.13   | 31.93 |
| Bottom-Up Summarization [Gehrmann et al., 2018] | 41.22   | 18.68   | 38.34   | 32.75 |
| DCA [Celikyilmaz et al., 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 et al., 2016]            | 42.2  | 24.9  |
| ML+RL+intra-attn [Paulus et al., 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.