

XMU' s Simultaneous Translation System at NAACL 2021

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Outline

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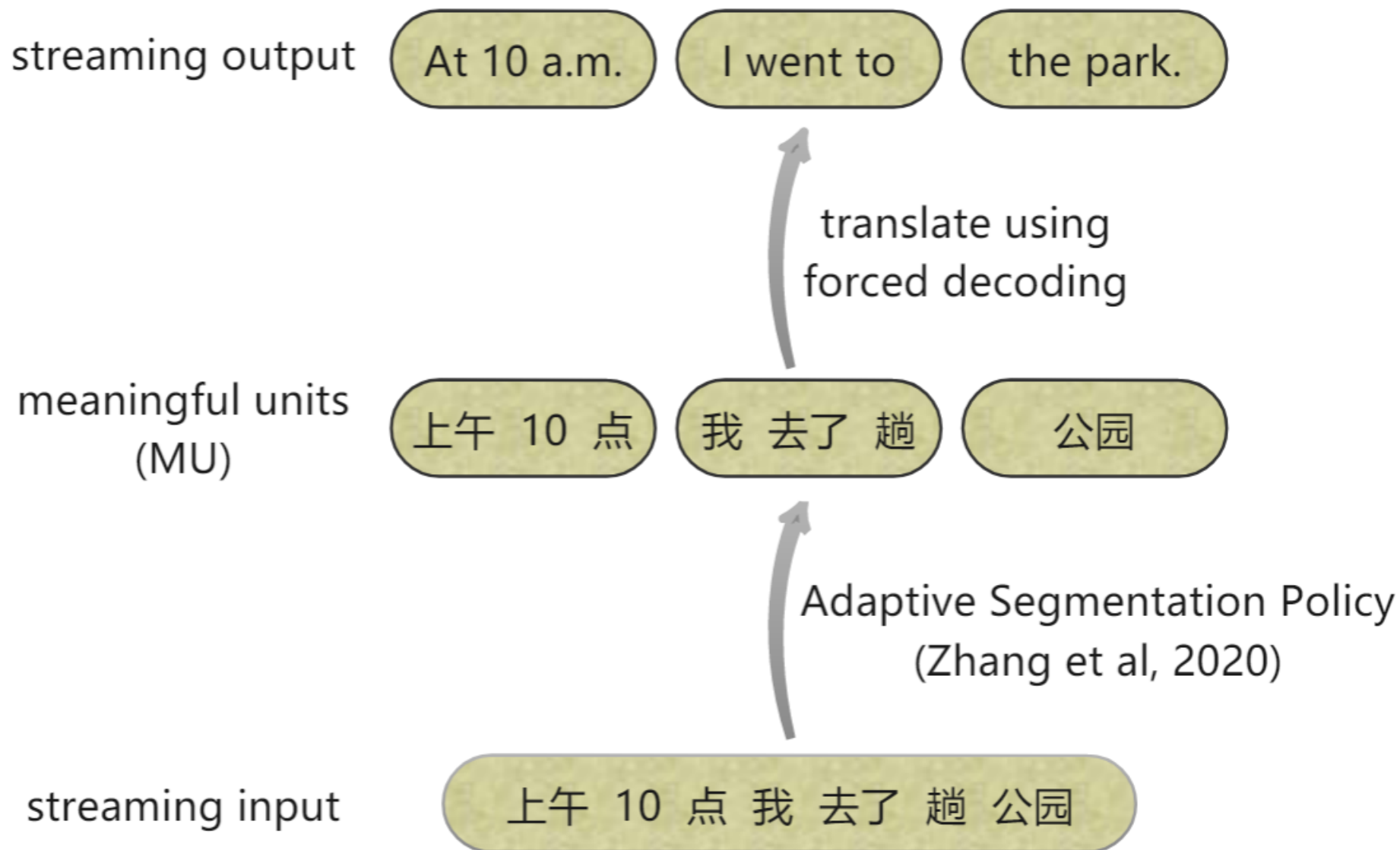
Tasks

- We participated in two tasks:
 - Zh->En Translation, input: streaming transcription. (text-to-text)
 - Zh->En Translation, input: audio file. (speech-to-text)

Data and preprocessing

- Datasets
 - Our MT model is pretrained on CWMT19 (9.1M parallel sentence pairs).
 - Our MT model is fine-tuned on Baidu Speech Translation Corpus (39K parallel sentence pairs).
- Preprocessing
 - Filter out long sentence pairs.
 - Convert full-width characters into half-width characters.
 - Segment Chinese text and tokenize English text.
 - Apply byte-pair-encoding to all sentences.

Text-to-text system



How the system translates streaming text.

Text-to-text system

- The MU segmentor is a text classifier based on BERT.
- Once the probability of class 1 is larger than a threshold δ , the input text is segmented.

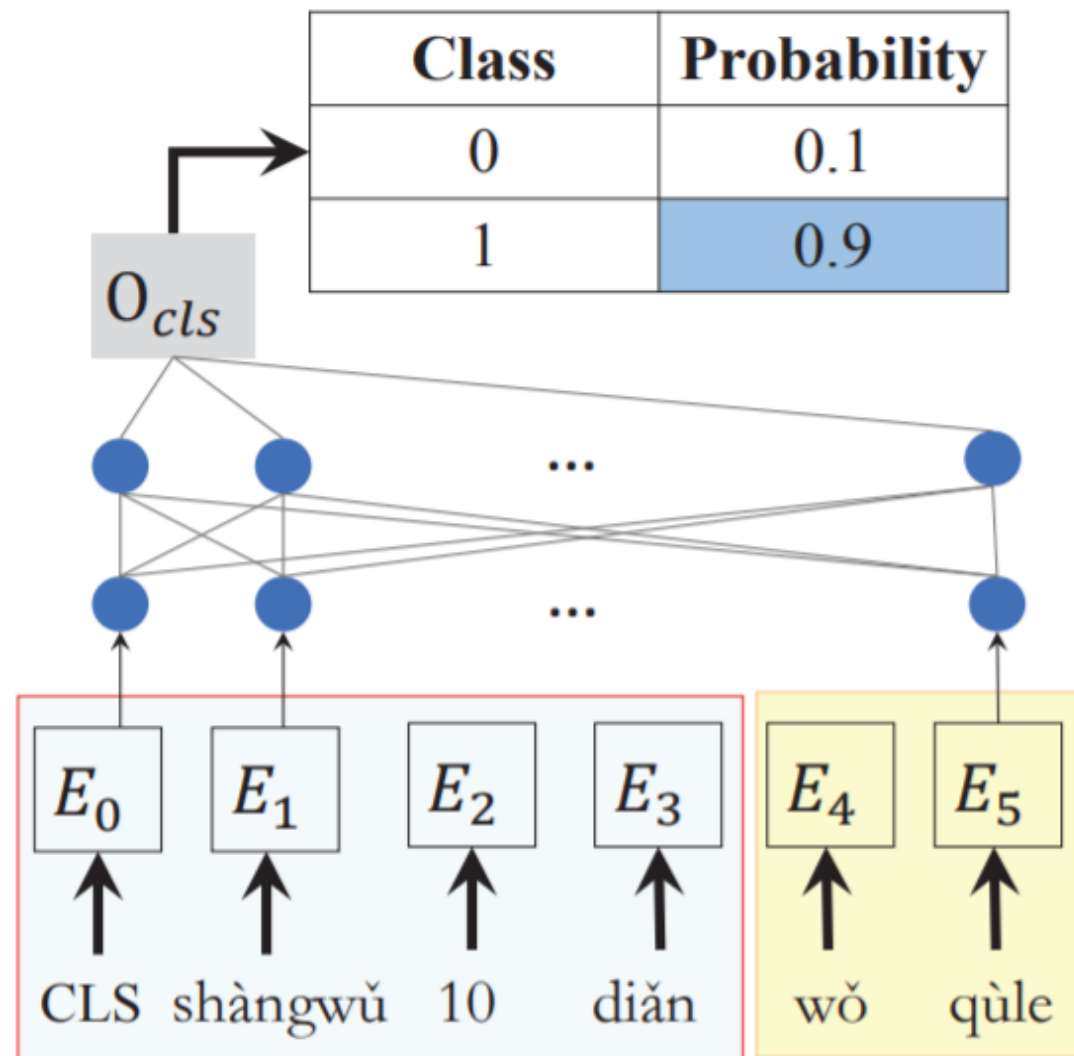


Illustration of the MU segmentor.

Text-to-text system

- To train the MU segmentor, we extract MUs and generate training data.

| t | c_t | $f_t(m = 2)$ | l_t |
|-----|------------------------------|---------------|-------|
| 1 | shàngwǔ | 10 diǎn | 0 |
| 2 | shàngwǔ 10 | diǎn wǒ | 0 |
| 3 | shàngwǔ 10 diǎn | wǒ qùle | 1 |
| 4 | shàngwǔ 10 diǎn wǒ | qùle tàng | 0 |
| 5 | shàngwǔ 10 diǎn wǒ qùle | tàng gōngyuán | 0 |
| 6 | shàngwǔ 10 diǎn wǒ qùle tàng | gōngyuán | 1 |

...

generate training examples

shàngwǔ 10 diǎn

wǒ qùle tàng

gōngyuán

extract MUs

shàngwǔ 10 diǎn wǒ qùle tàng gōngyuán

Generating examples for training the MU segmentor.

Text-to-text system

- Algorithm 1 extracts the MUs in an input sentence.
- We only use the *basic method* proposed by [Zhang et al, 2020](#) for extracting MUs.

Algorithm 1: Extract MUs

Input: $\mathbf{x} = x_1, \dots, x_T$ \triangleright streaming input

Output: \mathbf{S}_{MU} \triangleright list of MU segmentation

```
1  $k = 0$   $\triangleright$  position of last MU boundary
2  $\tilde{\mathbf{y}} = M_{\text{nmt}}(\text{src} = \mathbf{x}, \text{tgt}_{\text{force}} = \text{None})$ 
    $\triangleright$  full sentence decoding
3 while Reading  $x_t$  do
4      $\mathbf{y}^t = M_{\text{nmt}}(\text{src} = \mathbf{x}_{\leq t}, \text{tgt}_{\text{force}} = \mathbf{y}^k)$ 
5     if  $\mathbf{y}^t$  is a prefix of  $\tilde{\mathbf{y}}$  then
6          $\mathbf{S}_{\text{MU}} = \mathbf{S}_{\text{MU}} \cup \{x_{k+1}, \dots, x_t\}$ 
7          $k = t$ 
8 return  $\mathbf{S}_{\text{MU}}$ 
```

The algorithm for extracting MUs.

Speech-to-text system

- The speech-to-text system is a pipeline of three components:
 - Baidu's real time ASR service.
 - A repunctuation model, which is a BERT-based sequence labeling model.
 - A text-to-text translation subsystem.

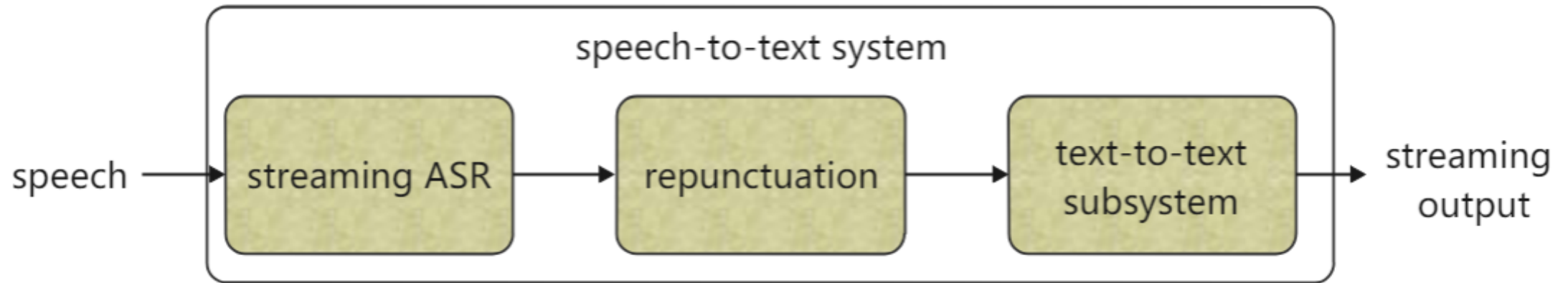
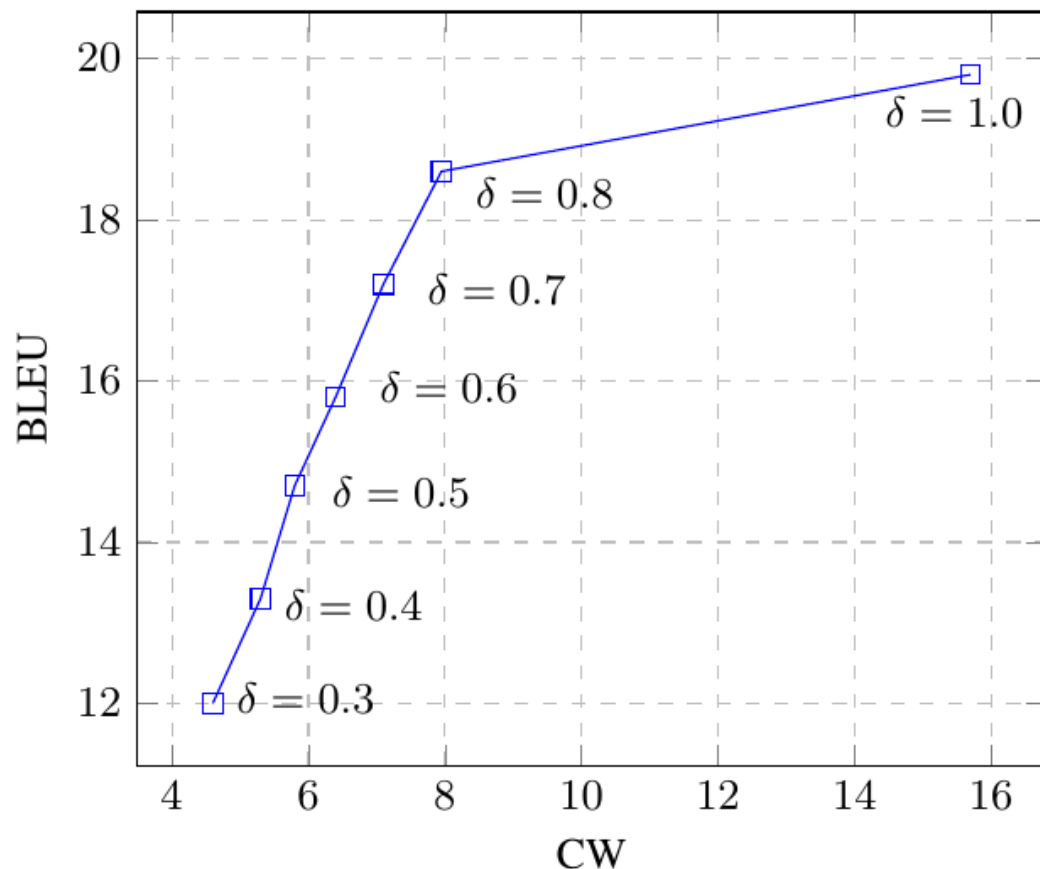


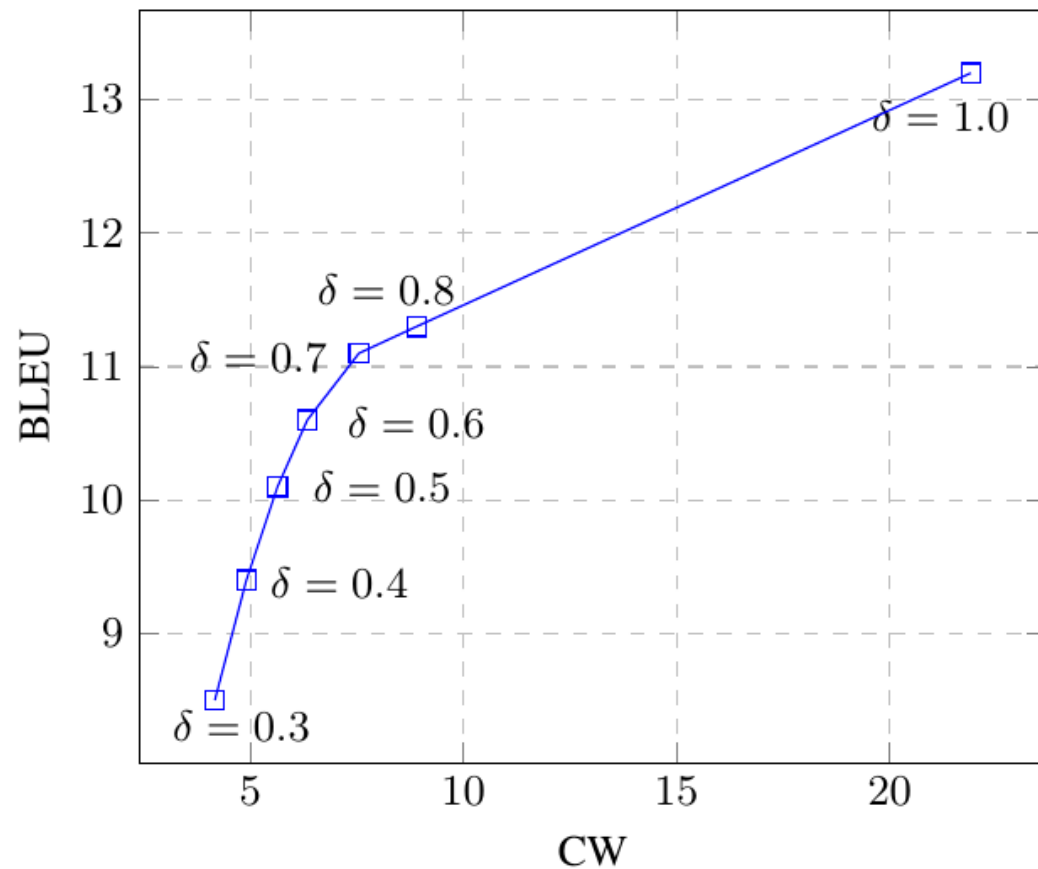
Illustration of speech-to-text system.

Experiments

- The two systems are evaluated on the development set of the Baidu Speech Translation Corpus.



The text-to-text system.



The speech-to-text system.

Conclusion and future work

- Adaptive Segmentation Policy (Zhang et al., 2021) is effective.
- Our systems use a conventional MT model that is not designed for simultaneous translation. We will study how to train the MT model that is more suitable for simultaneous translation.