

# NAIST Simultaneous Speech-to-Text Translation System for IWSLT 2022

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**Abstract** End-to-end SimulST using adaptive segmentation policies based on bilingual prefix alignment [Kano et al., 2022]

## Bilingual Prefix Alignment (BPA) for SimulST

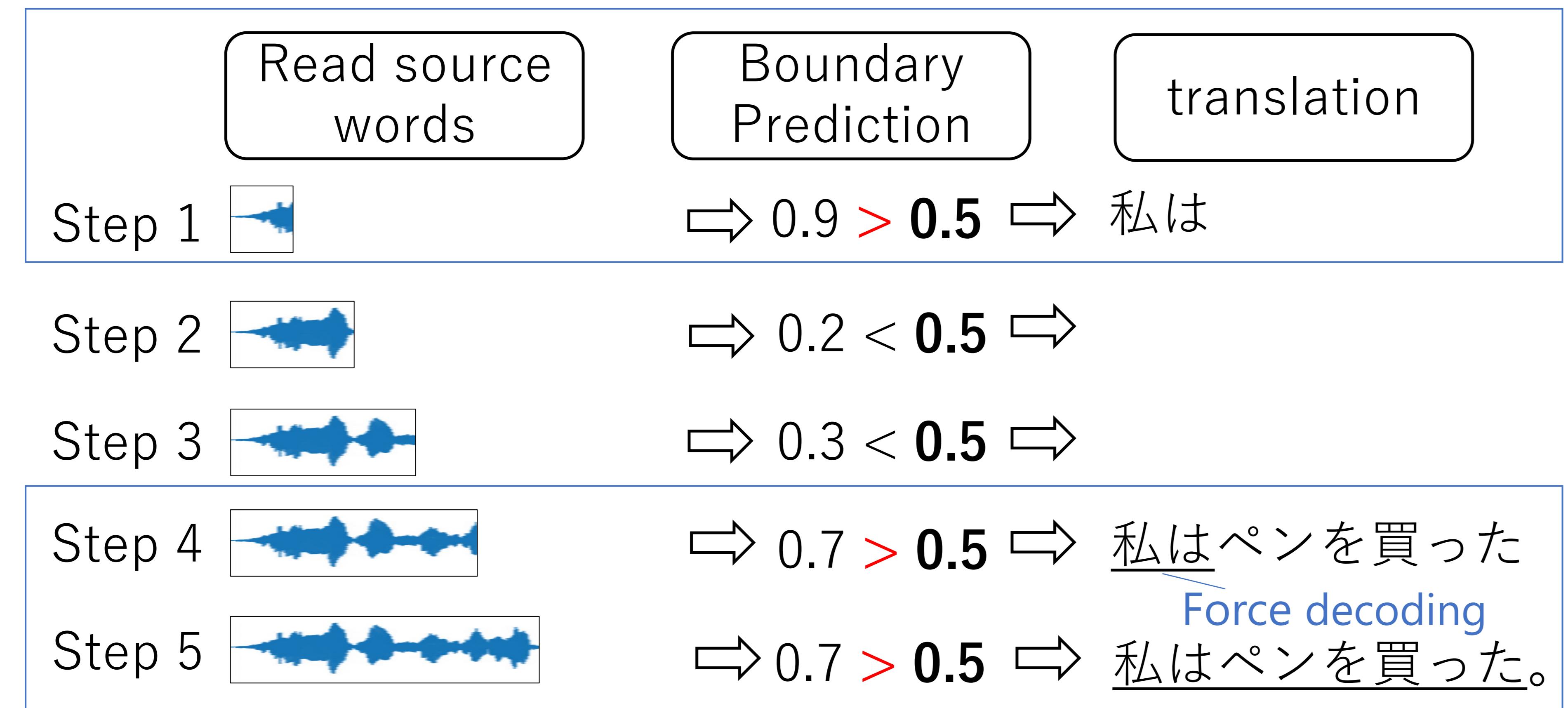
### ● Training process

- (1) Fine-tune (FT) offline ST
- (2) Boundary Predictor (BP)

from bilingual prefix pairs data

### ● Translation process

- BP detects segment boundaries
- ST translates a partial input taking translation history into account



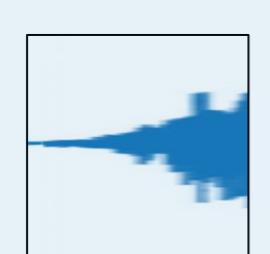
**Full-sentence Output** 私は ペンを 買った

*common target prefix*

**Part Output** 私は

*Extracting prefix pairs*

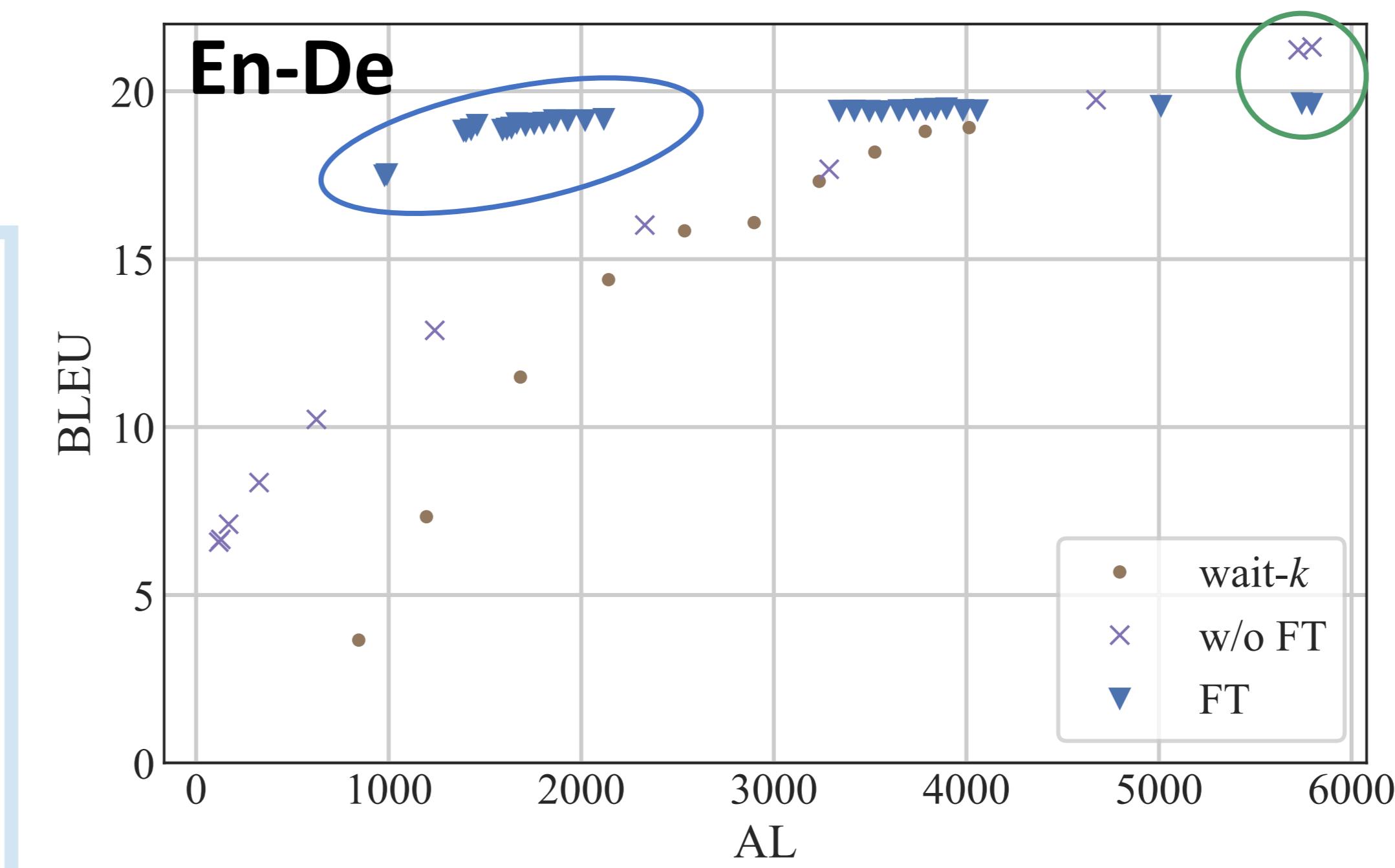
**Offline ST**



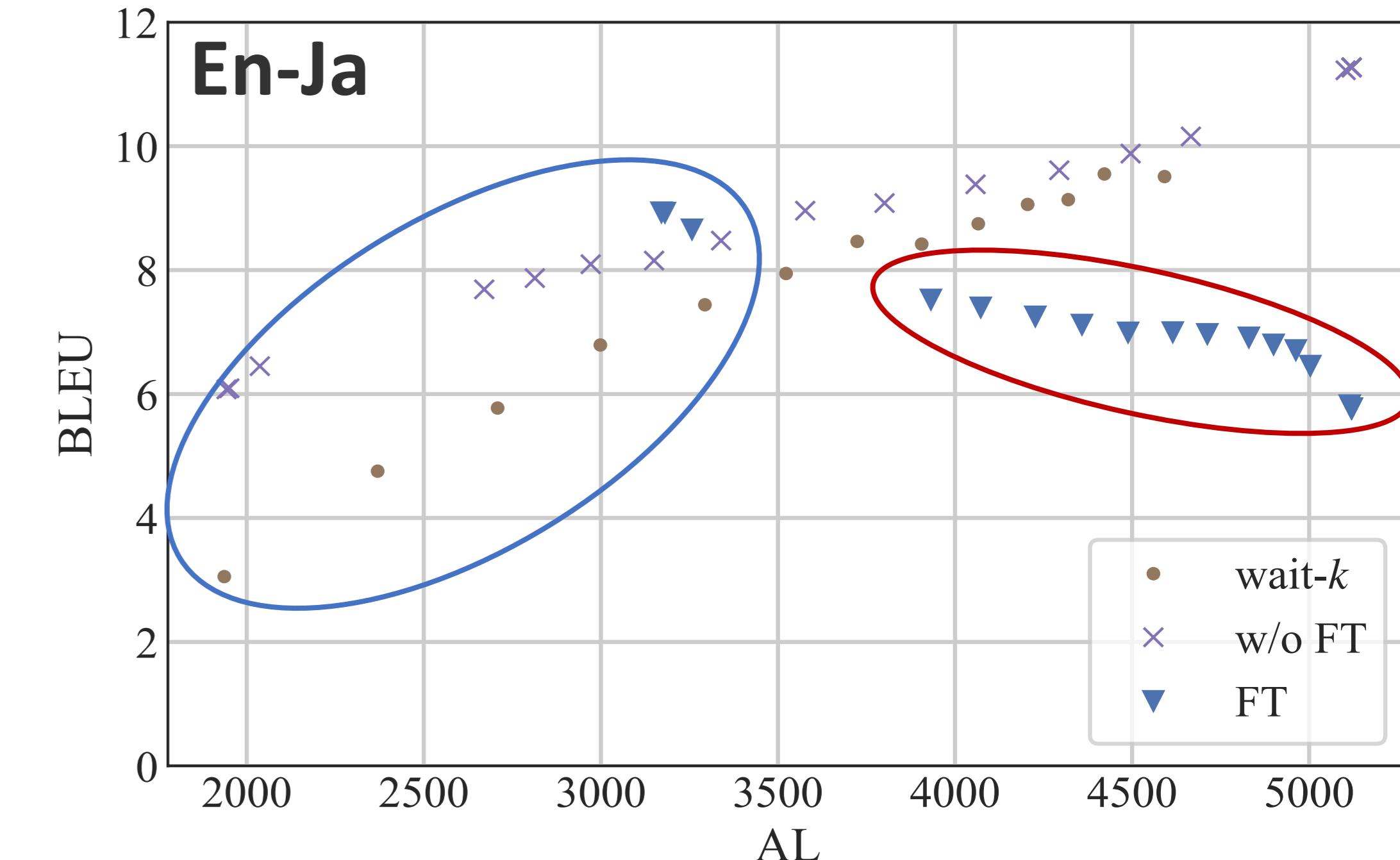
- (1) FT data  
[私, 私は]  
[私は, 私は ペンを 買つ]
- (2) BP data  
[私, 1..1]  
[私, 1..10..0]  
[私, 110..00..0]  
[私, 1..10..00..01..1]  
[私, 1..10..00..01..11..1]

**Experiment Data MuST-C v2 En-De, En-Ja ST Model Transformer Boundary Prediction Model LSTM, 100 frames/unit**

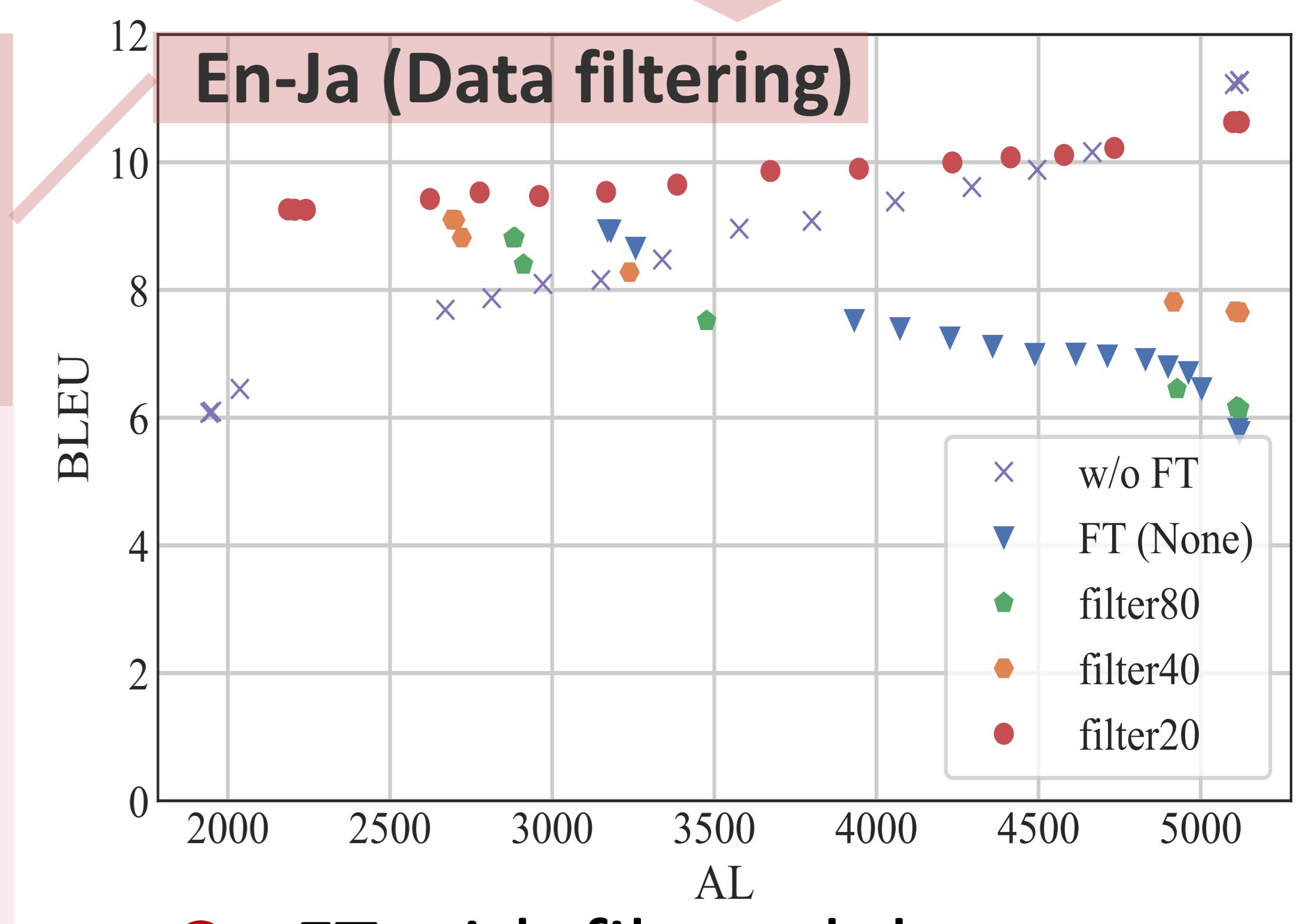
**Evaluation** MuST-C v2 tst-COMMON in SimulEval



- FT > w/o FT in AL<=4000  
→ Robust to lower latency
- w/o FT > FT in high latency



- w/o FT > wait-k
- FT < wait-k, w/o FT  
→ prefer too short outputs



- FT with filtered data achieved best performance!

**Conclusion** Our results show effectiveness of fine-tuned ST & Boundary Predictor with bilingual prefix alignment

- Robust to lower latency
- En-Ja : data filtering of large length gap pairs was effective due to its sentence structure