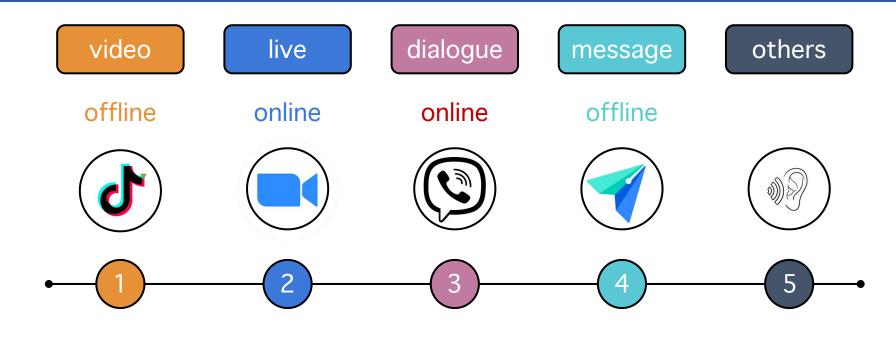


# 国际人工智能会议 AAAI 2021 论文北京预讲会

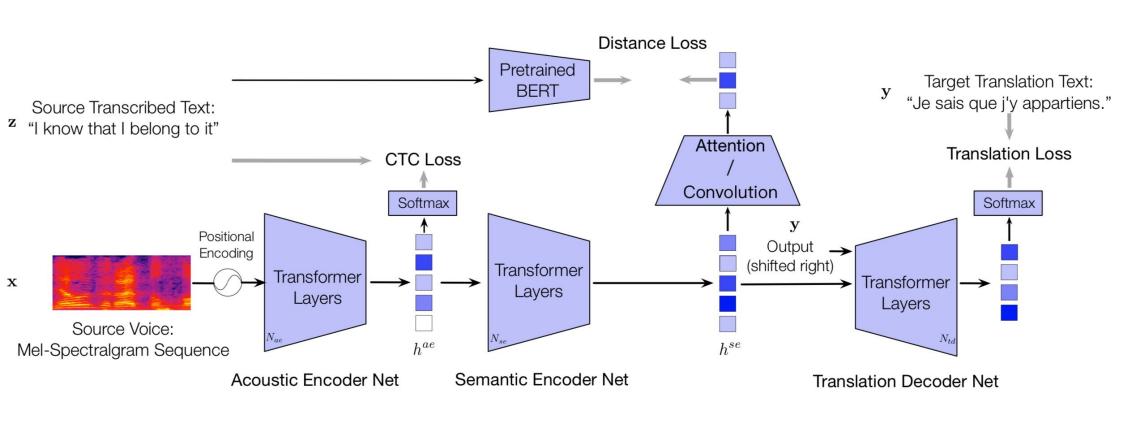
## 多人 火山翻译

#### 1 Background

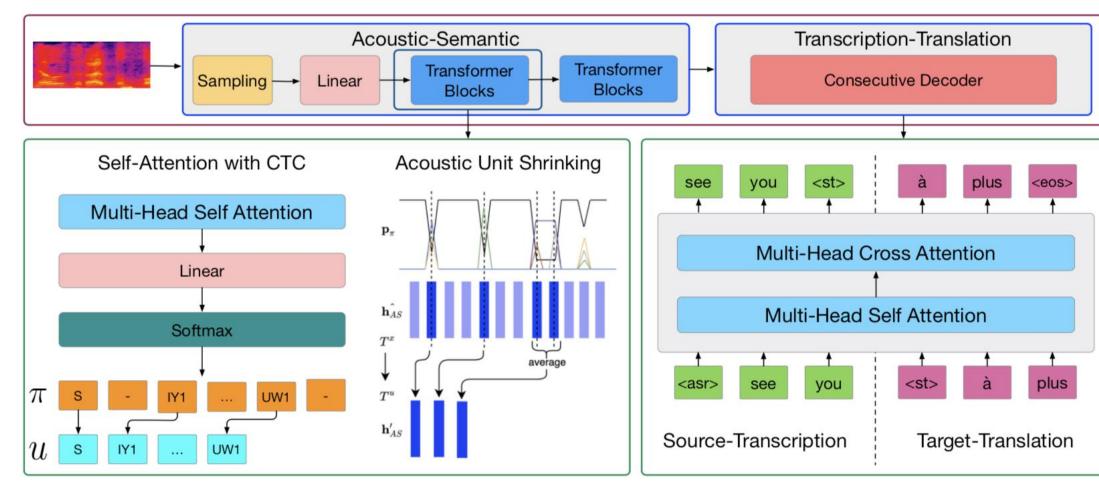


# Cascaded System : End-to-end System: Hello! ASR MT 안녕!

#### 2 Listen, Understand and Translate



## 3 Consecutive Decoding



### 2.1 Methodology

Our proposed LUT consists of three modules, including an acoustic encoder, a semantic encoder and a translation decoder:

- ☐ An acoustic encoder network that encodes the audio input sequence into hidden features corresponding to the source text;
- ☐ A semantic encoder network that extracts hidden semantic representation for translation, which behaves like a normal machine translation encoder;
- ☐ A translation decoder network that emits sentence tokens in the target language.

#### 3.1 Methodology

We divide our method COSTT into two phases, including the acoustic-semantic modeling phase (AS) and the transcription-translation modeling phase (TT).

- ☐ The AS phase accepts the speech features, outputs the acoustic representation, and encodes the shrunk acoustic representation into semantic representation.
- ☐ The TT phase accepts the AS's representation and consecutively outputs source transcription and target translation text sequences with a single shared decoder.

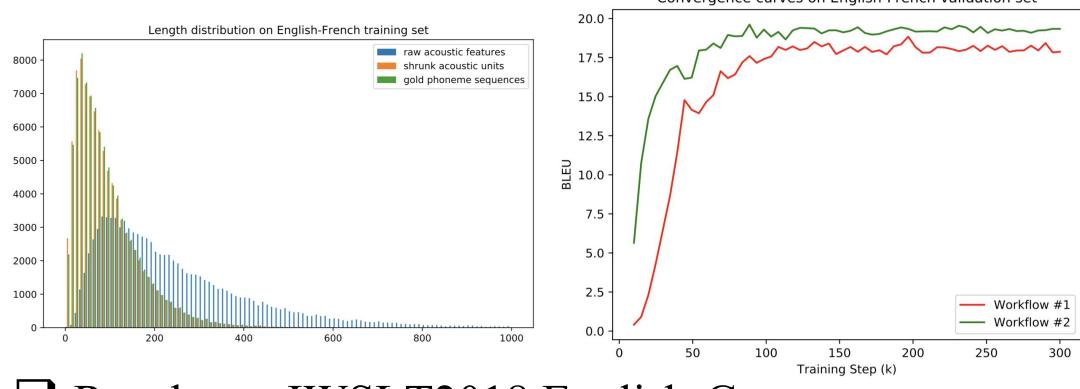
#### 2.2 Experiments

#### ☐ Results on Librispeech English-French

Method	Enc Pre-train (speech data)	Dec Pre-train (text data)	greedy	beam
MT system				
Transformer MT	-	-	20.98	21.51
Base ST setting				
LSTM ST (Bérard et al. 2018)	X	×	12.30	12.90
+pre-train+multitask (Bérard et al. 2018)	✓	✓	12.60	13.40
LSTM ST+pre-train (Inaguma et al. 2020)	✓	✓	-	16.68
Transformer+pre-train (Liu et al. 2019a)	✓	✓	13.89	14.30
+knowledge distillation (Liu et al. 2019a)	✓	✓	14.96	17.02
TCEN-LSTM (Wang et al. 2019)	✓	✓	-	17.05
Transformer+ASR pre-train (Wang et al. 2020)	✓	X	-	15.97
Transformer+curriculum pre-train (Wang et al. 2020)	✓	X	:=:::	17.66
LUT	×	×	16.70	17.75
Expanded ST setting				
LSTM+pre-train+SpecAugment (Bahar et al. 2019)	<b>√</b> (236h)	✓	-	17.00
Multilingual ST+PT (Inaguma et al. 2019)	<b>√</b> (472h)	X	-	17.60
Transformer+ASR pre-train (Wang et al. 2020)	<b>√</b> (960h)	X		16.90
Transformer+curriculum pre-train (Wang et al. 2020)	<b>√</b> (960h)	X	_	18.01
LUT	<b>√</b> (207h)	×	17.55	18.34

#### 3.2 Experiments

#### ☐ Effects of Shrinking Mechanism & Pre-training



#### ☐ Results on TED English-Chinese

Method	Enc Pre-train (speech data)	Dec Pre-train (text data)	BLEU
MT system			
Transformer MT (Liu et al. 2019a)	-	-	27.08
Base setting			
Transformer+pre-train (Liu et al. 2019a)	✓	✓	16.80
+knowledge distillation (Liu et al. 2019a)	✓	✓	19.55
Multi-task+pre-train* (Inaguma et al. 2019)(re-implemented)	✓	X	20.45
LUT	X	X	20.84

#### ☐ Results on IWSLT2018 English-German

Method	(speech data)	Dec Pre-train (text data)	tst2013
MT system			
Transformer MT	-	-	27.87
Base setting			
ESPnet (Inaguma et al. 2020)	×	X	12.50
+enc pre-train	✓	X	13.12
+enc dec pre-train	✓	✓	13.54
Transformer+ASR pre-train (Wang et al. 2020)	✓	X	15.35
+curriculum pre-train (Wang et al. 2020)	✓	X	16.27
COSTT	×	×	16.30
Expanded setting			
Multi-task+pre-train (Inaguma et al. 2019)	<b>√</b> (472h)	X	14.60
CL-fast* (Kano, Sakti, and Nakamura 2018)	<b>√</b> (479h)	X	14.33
TCEN-LSTM (Wang et al. 2019)	<b>√</b> (479h)	<b>√</b> (40M)	17.67
Transformer+curriculum pre-train (Wang et al. 2020)	<b>√</b> (479h)	<b>√</b> (4M)	18.15
COSTT	<b>√</b> (272h)	<b>√</b> (1M)	18.63

主办方: 中国中文信息学会青年工作委员会

承办方: 智源社区



