

## **Towards Language-Universal Mandarin-English Speech Recognition Using DFSMN-CTC-sMBR**

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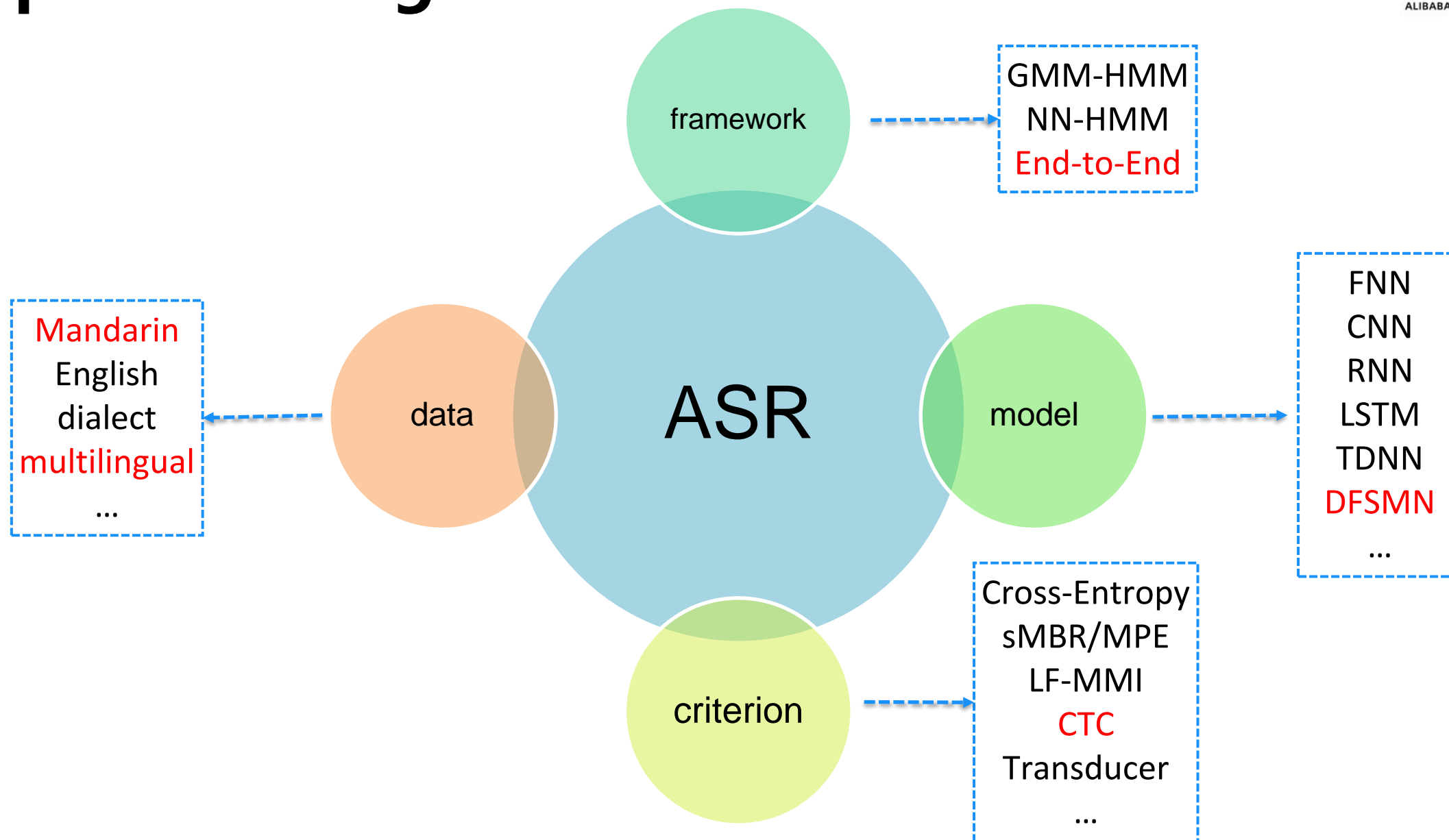
## ■ Mandarin Speech Recognition using DFSMN-CTC-sMBR

- Zhang S., Lei M., Acoustic Modeling with DFSMN-CTC and Joint CTC-CE Learning. In *Interspeech* 2018 (pp. 771-775).
- Zhang S., Lei M., Liu Y, et al. , Investigation of Modeling Units for Mandarin Speech Recognition Using DFSMN-CTC-sMBR. Proc. of ICASSP 2019: 7085-7089.

## ■ Language-Universal Mandarin-English Speech Recognition

- Zhang, S., Liu, Y., Lei, M., Ma, B., Xie, L. , Towards Language-Universal Mandarin-English Speech Recognition. Proc. of Interspeech 2019, 2170-2174.

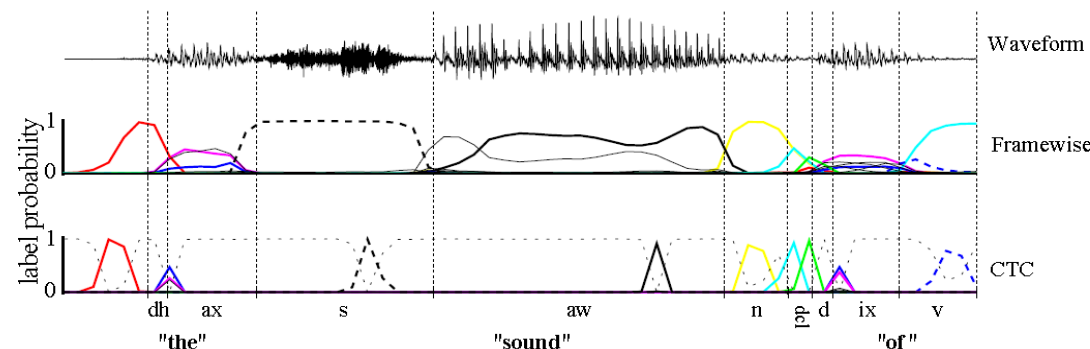
# Speech Recognition



# CTC based Acoustic Modeling

## ■ Connectionist Temporal Classification

- First proposed by Grave @2006 @2013
- Successfully applied to LVCSR by Google @2016
- LSTM/BLSTM-CTC



## ■ CTC Vs. Cross-Entropy

- Advantage: better performance, **faster decoding speed**
- Problem: unstable, spike delay et al.

## ■ CTC for Mandarin speech recognition

- Deep Speech 2 【2016】 , Zhehuai Chen 【2016】 , Zhongdi Qu 【2017】 et al.
- Zhang S., Lei M., Acoustic Modeling with DFSMN-CTC and Joint CTC-CE Learning. In *Interspeech* 2018 (pp. 771-775).
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# Mandarin Speech Recognition with DFSMN-CTC-sMBR

## ■ Acoustic Modeling units for Mandarin

序列(Sequence)	• 玩具总动员	数目无穷
词(word)	• 玩具 总动员	上百万
字(character)	• 玩 具 总 动 员	数十万, 常用5千
音节(syllable)	• wan2 ju4 zong3 dong4 yuan2	一千多
绑定的声韵母(CD-IF)	• /s-w-an2 w-an2-j an2-j-u4 ...	决策树, 数千
声韵母(IF)	• w an2 j u4 z ong3 d ong4 y uan2	143个

- 传统语音识别: 绑定的声韵母
- 基于CTC的识别模型: 声韵母、绑定的声韵母、音节、**字**

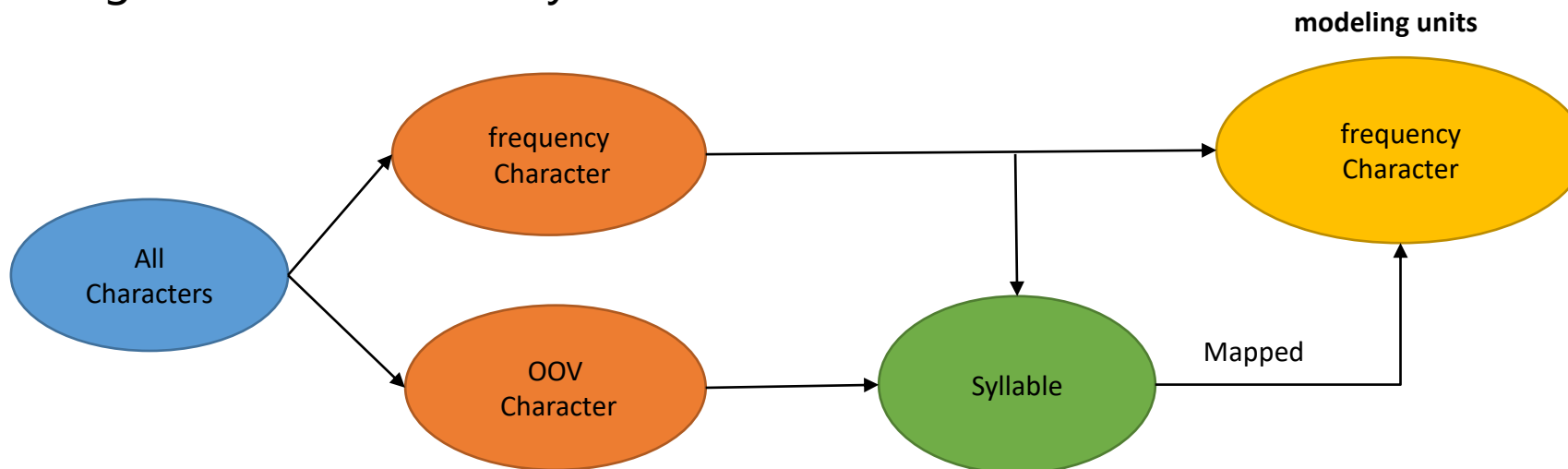
# Mandarin Speech Recognition with DFSMN-CTC-sMBR

## ■ Modeling units: Mixed-syllable-character

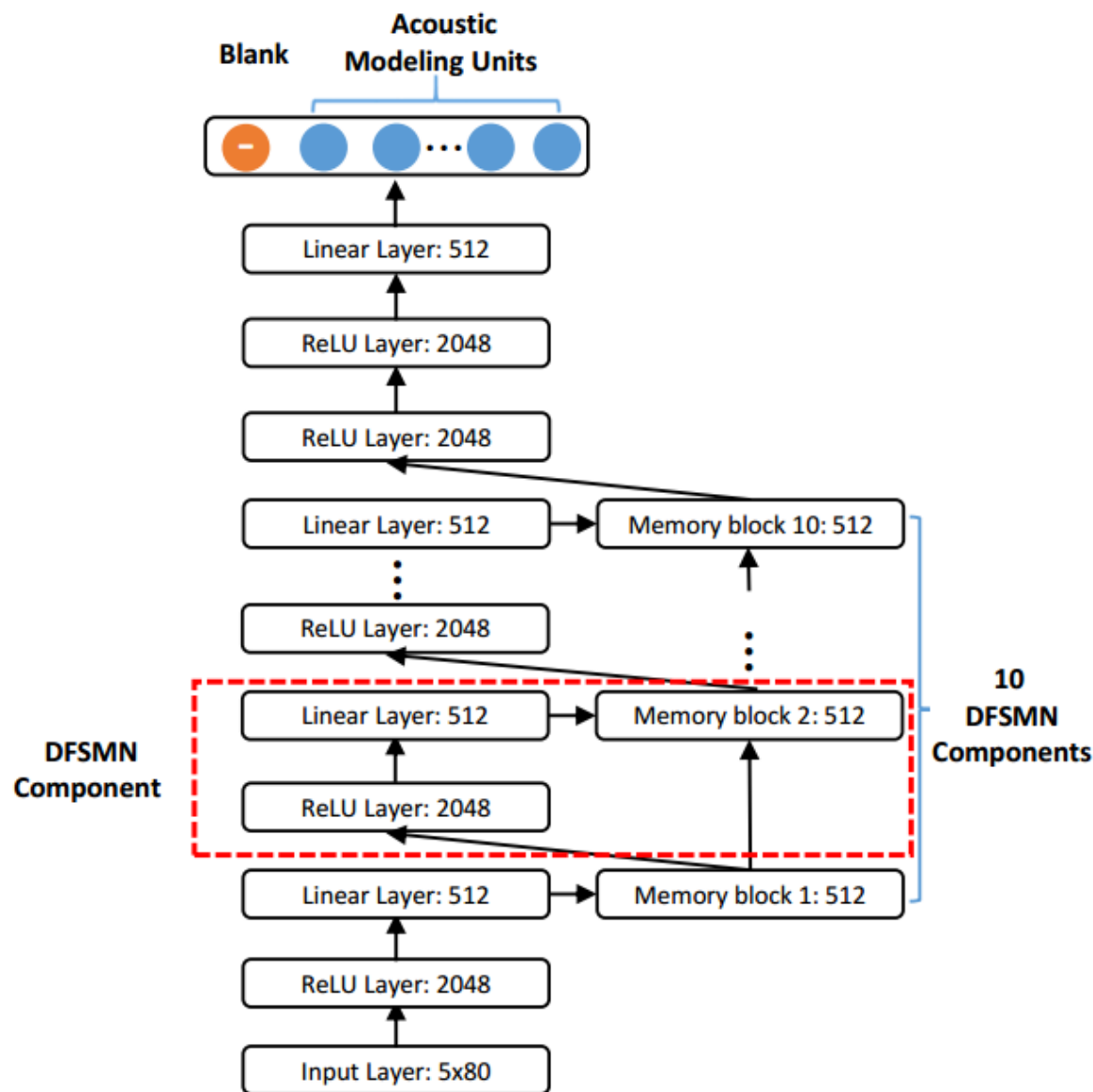
- Method: frequency character + syllable
- Advantage: without OOV problem; lower frame rate

## ■ Modeling units: All-Character

- Method: frequency character+ character mapping
- Advantage: small vocabulary size



# Mandarin Speech Recognition with DFSMN-CTC-sMBR



## Why DFSMN ?

- Feedforward architecture
- Effectively model long-term dependency

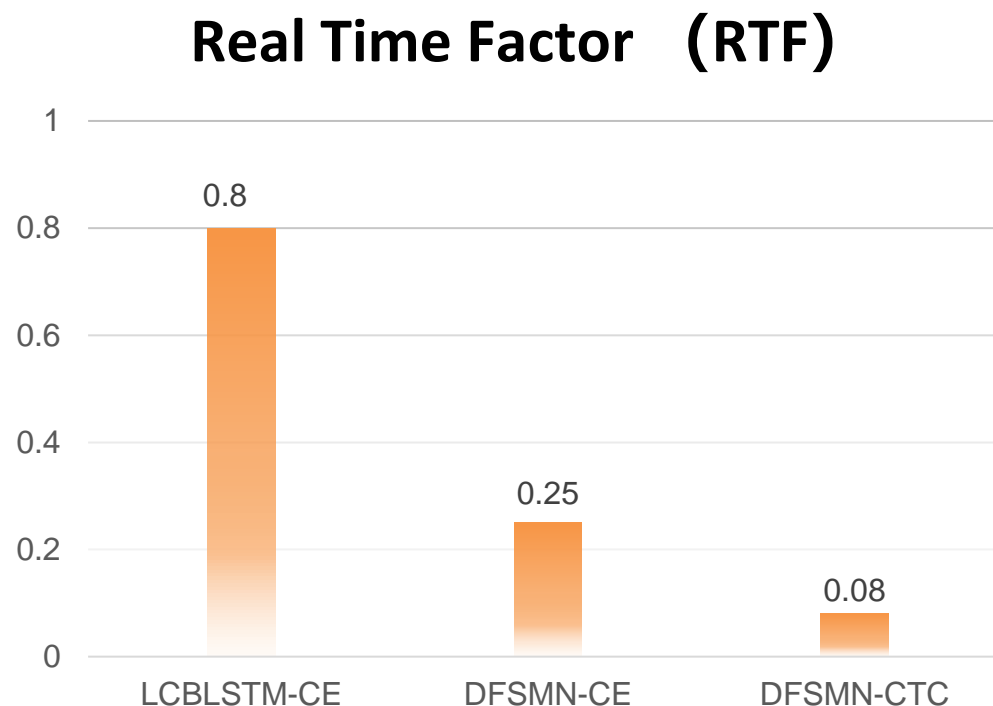
## Acoustic Modeling units

Modeling units	Detailed composition
CI-IF	23 Initials + 185 tonal Finals
CD-IF	7951 context-dependent Initial/Finals
Syllable	1319 tonal syllables
Char(2k)-Syllable	2000 high frequency Chinese characters + 1319 tonal syllables
Char(3k)-Syllable	3000 high frequency Chinese characters + 1319 tonal syllables
All-Char(2k)	2000 high frequency Chinese characters

# DFSMN-CTC-sMBR for Mandarin Speech Recognition

## ■ 20000-hours-task

Exp	Model	Modeling Units	CER(%)	
			CE	+sMBR
1	LCBLSTM	CD-IF	11.32	10.59
2	DFSMN(10)	CD-IF	10.53	9.49
			CTC	+sMBR
3	DFSMN(10)	CI-IF	10.38	9.37
4	DFSMN(10)	CD-IF	9.70	-
5	DFSMN(10)	Syllable	9.03	7.94
6	DFSMN(10)	Char(2k)+Syllable	8.87	7.61
7	DFSMN(10)	Char(3k)+Syllable	8.81	7.45
8	DFSMN(12)	Char(3k)+Syllable	8.46	7.28
9	DFSMN(10)	All-char(2k)	8.08	6.98





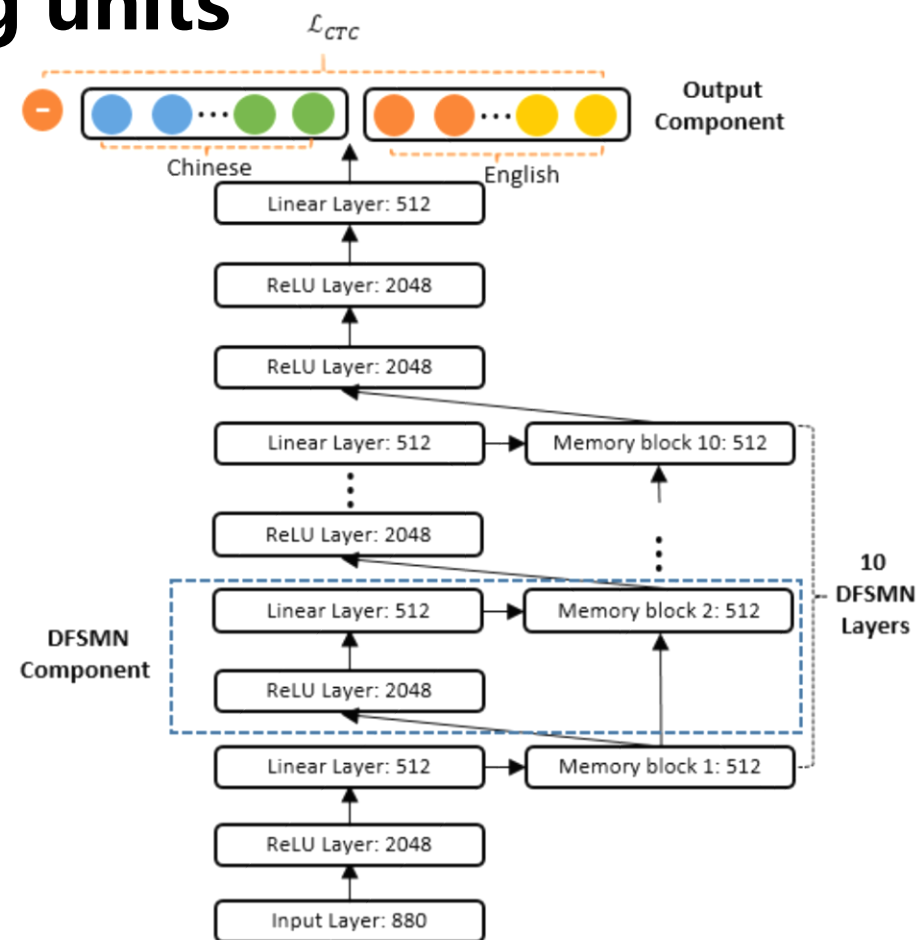
# Language-Universal Mandarin-English ASR

- Background: language-specific ASR system
- Language-Universal Mandarin-English ASR System
  - ASR system can recognize Mandarin, English and code-switching speech without any language-specific information.
- Challenge
  - How to model two languages
  - Code-switching acoustic & text data
  - Discriminative acoustic score
- Our works
  - CTC-sMBR
  - Bilingual-AM
  - Word Space Mapping

# Language-Universal Mandarin-English ASR

## ■ Idea 1、Mixed Acoustic Modeling units

- Acoustic Model
  - DFSMN-CTC-sMBR
- Training Data
  - monolingual Mandarin + English
- Acoustic Modeling Units
  - Mandarin: All-character
  - English: wordpiece



# Language-Universal Mandarin-English ASR

## ■ Idea 1、Mixed Acoustic Modeling units

- Training data
  - 20000-hours-Mandarin + 15000 hours English

System	Criterion	Mand.(CER%)	Eng.(WER%)
Mandarin	CTC	8.08	-
	+SMBR	<b>6.98</b>	-
English	CTC	-	13.47
	+SMBR	-	<b>11.31</b>

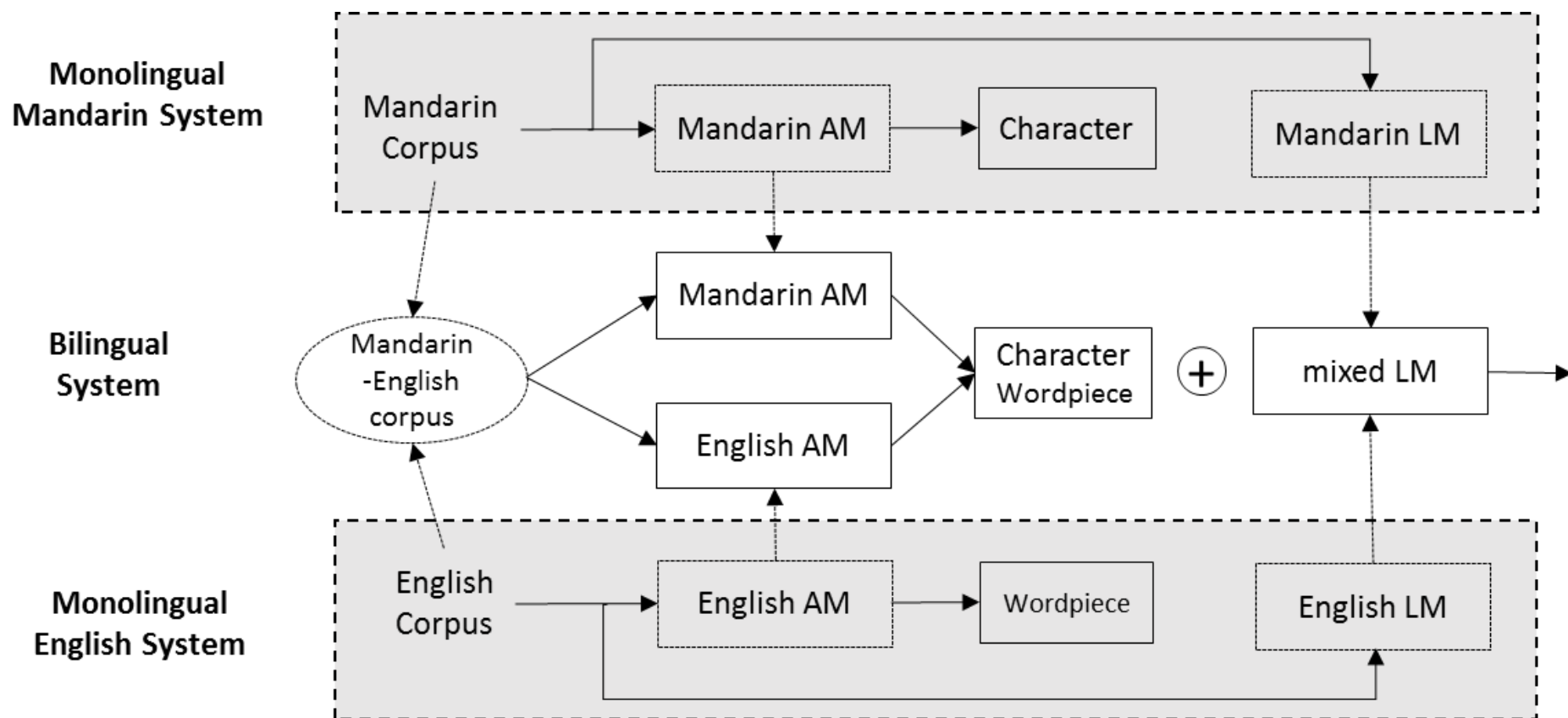
Language-specific DFSMN-CTC-sMBR system

ID	Data		Lang. Info.	Mand. (CER%)	Eng. (WER%)
	Mandarin	English			
1	100%	10%	w	8.28	22.61
			w/o	8.44	23.23
2	100%	50%	w	9.08	16.44
			w/o	9.60	17.01
3	100%	100%	w	<b>9.52</b>	<b>15.04</b>
			w/o	9.83	15.66

Language-universal DFSMN-CTC system with mixed modeling units

# Language-Universal Mandarin-English ASR

## ■ Idea2、 Monolingual pre-trained Bilingual-AM



# Language-Universal Mandarin-English ASR

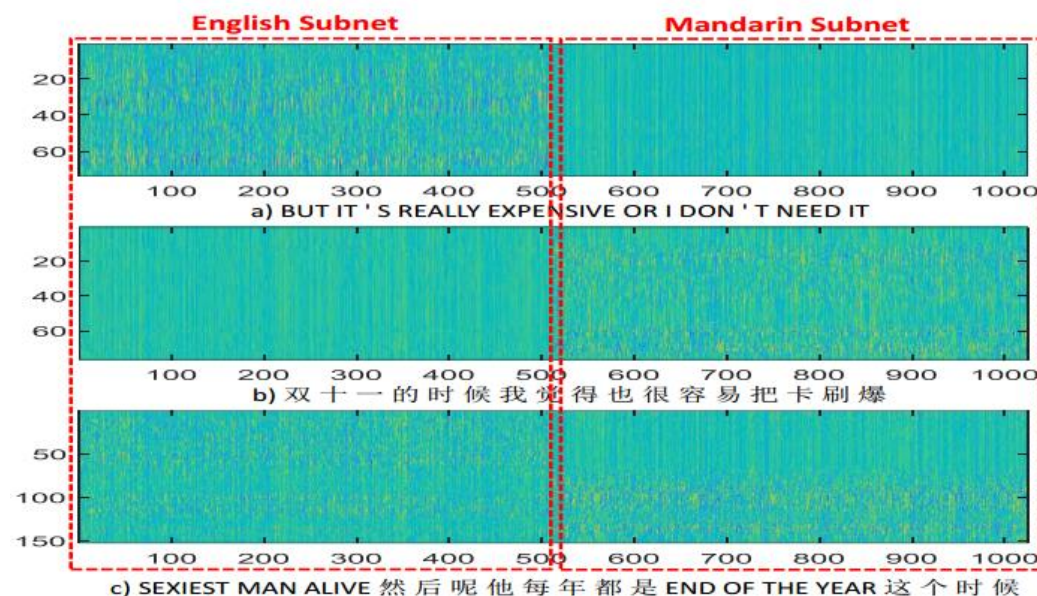
## ■ Idea2、 Monolingual pre-trained Bilingual-AM

Language-specific DFSMN-CTC-sMBR system

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	+SMBR	-	<b>11.31</b>

Language-universal DFSMN-CTC-sMBR system

Lang. Info.	Criterion	Mand.(CER%)	Eng.(WER%)
w	CTC	8.04	12.80
	+SMBR	<b>6.94</b>	<b>11.33</b>
w/o	CTC	8.14	12.94
	+SMBR	<b>7.02</b>	<b>11.60</b>



# Language-Universal Mandarin-English ASR

- Idea2、 Monolingual pre-trained Bilingual-AM
  - Code-switching test set

System	LM	Code-Switching Test Set		
		M	E	All
Baseline Bilingual	3-gram	8.73	20.74	<b>9.51</b>
	1-gram	10.14	21.71	10.89
Proposed Bilingual	3-gram	7.70	17.03	<b>8.31</b>
	1-gram	9.03	17.82	9.60

- Code-switching LM score is mostly back-off to 1-gram

# Language-Universal Mandarin-English ASR

## ■ Idea3、Word Space Mapping

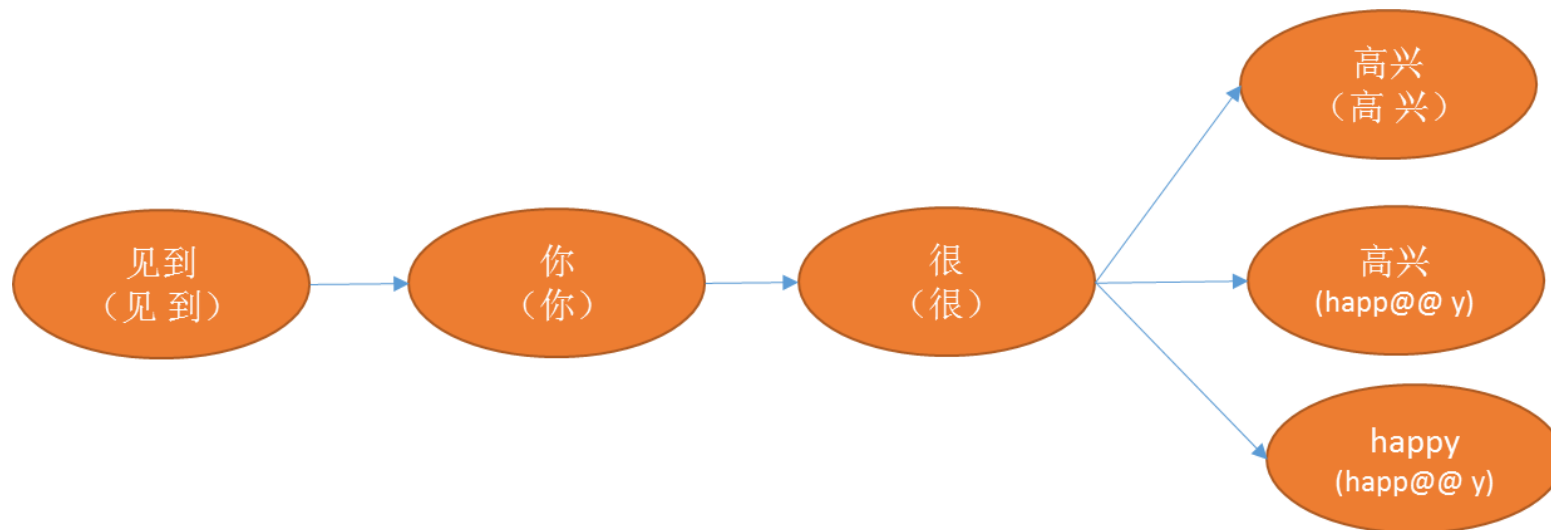
Case: 见到你很 happy

LM score:  $P(\text{happy} | \text{你很}) \rightarrow P(\text{happy})$

Word Mapping:  $f(\text{happy}) \rightarrow \text{高兴}$ ,  $P(\text{happy} | \text{你很}) \rightarrow P(\text{高兴} | \text{你很})$

Lexicon:

词条	发音
高兴	高兴
高兴	happ@@ y
happy	happ@@ y



# Language-Universal Mandarin-English ASR

## ■ Idea3、 Word Space Mapping

Method	Test1	Test2
Base	17.12(33.40)	18.54(25.15)
<b>WSP</b>	<b>16.98[32.73]</b>	<b>17.56[23.63]</b>

Label	处 理 data 的 部分
Base	true data 部 分
LM-Mapping	处 理 data 【数据】 部 分
Label	那 这 个 machine learning 我 会 主 要 觉 得 它 是 一 个 三 块 三 块 这 个 knowledge 的 结 合 哦
Base	那 这 个 machine learning 我 会 主 要 觉 得 他 是 一 个 三 块 三 块 这 个 knowledge that 结 合
LM-Mapping	那 这 个 machine learning 我 会 主 要 觉 得 他 是 一 个 三 块 三 块 这 个 knowledge 【知识】 的 结 合



# Language-Universal Mandarin-English ASR

## ■ Demo

达摩院  机器智能技术

thanks 