

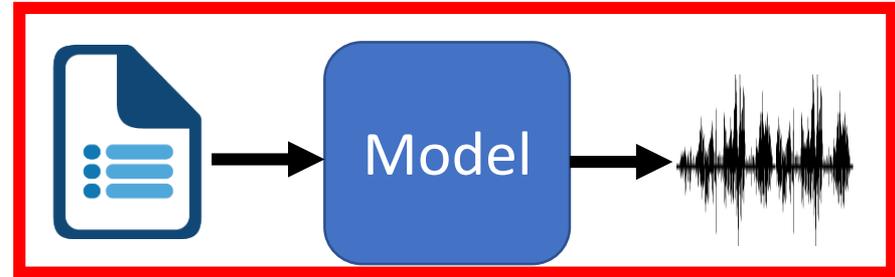
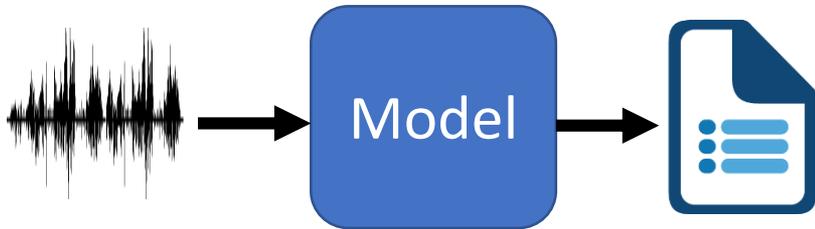


Speech Synthesis

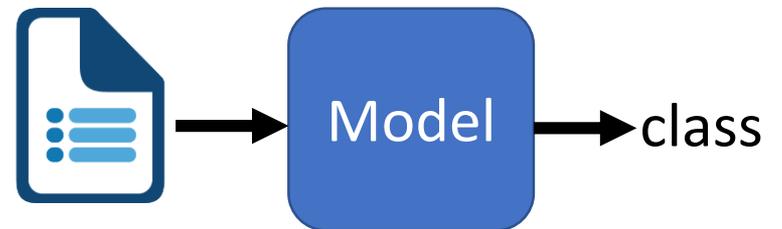
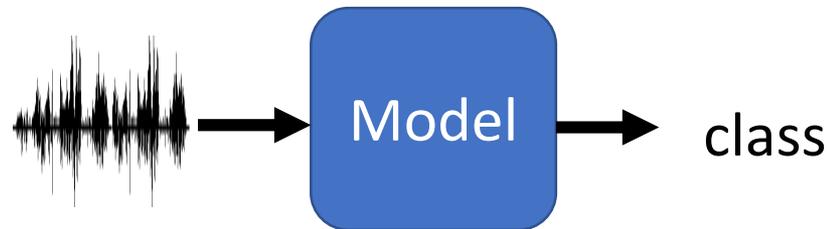
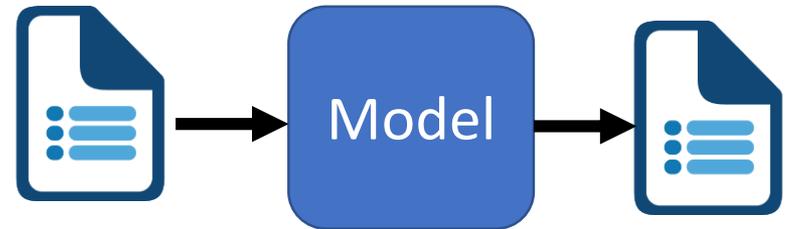
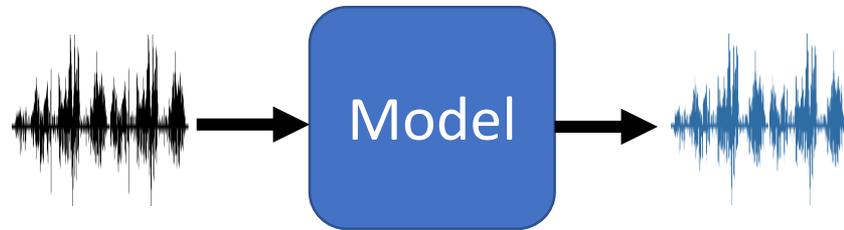
李宏毅

Hung-yi Lee

One slide for this course



Text-to-Speech (TTS) Synthesis



Outline

TTS before End-to-end

Tacotron: End-to-end TTS

Beyond Tacotron

Controllable TTS

VODER (1939)

<https://en.wikipedia.org/wiki/Voder>



Source of video: <https://www.youtube.com/watch?v=0rAyrmm7vv0>

IBM computer (1960s)

- In 1961, John Larry Kelly Jr. using an IBM computer to synthesize speech at Bell lab.



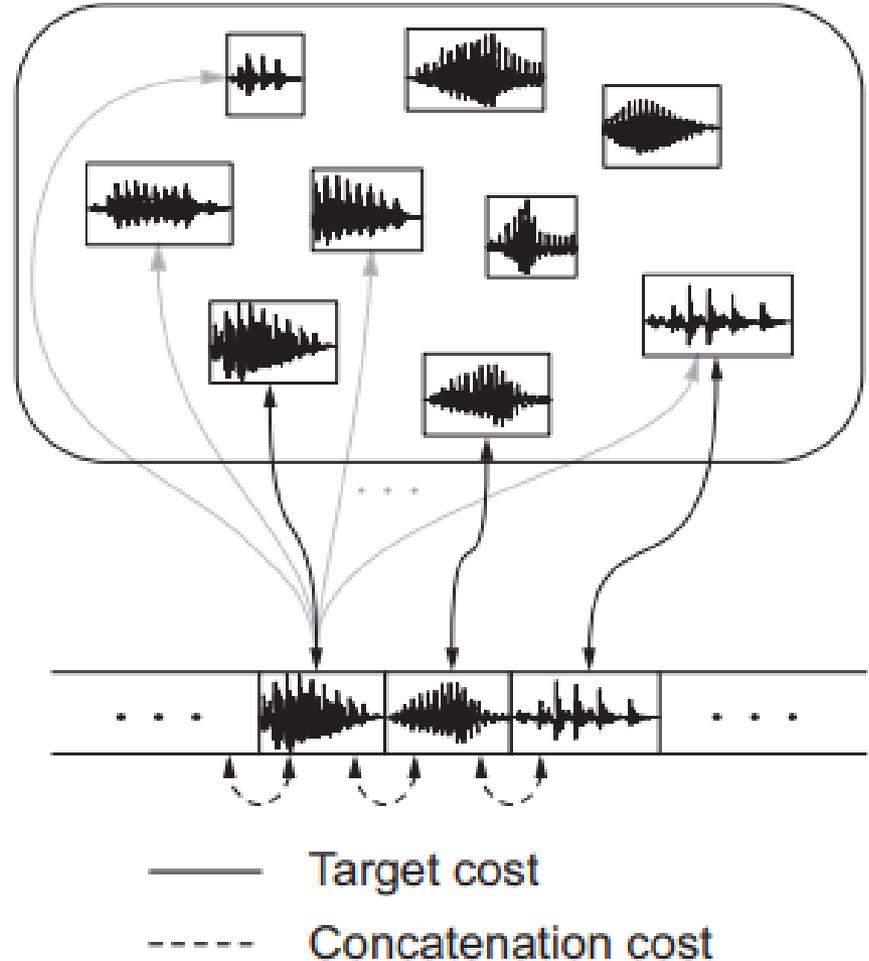
Source of video and audio: <https://youtu.be/UGsfwhb4-bQ>

https://www.vintagecomputermusic.com/mp3/s2t9_Computer_Speech_Demonstration.mp3

Concatenative Approach

speeches from a
large database

All segments

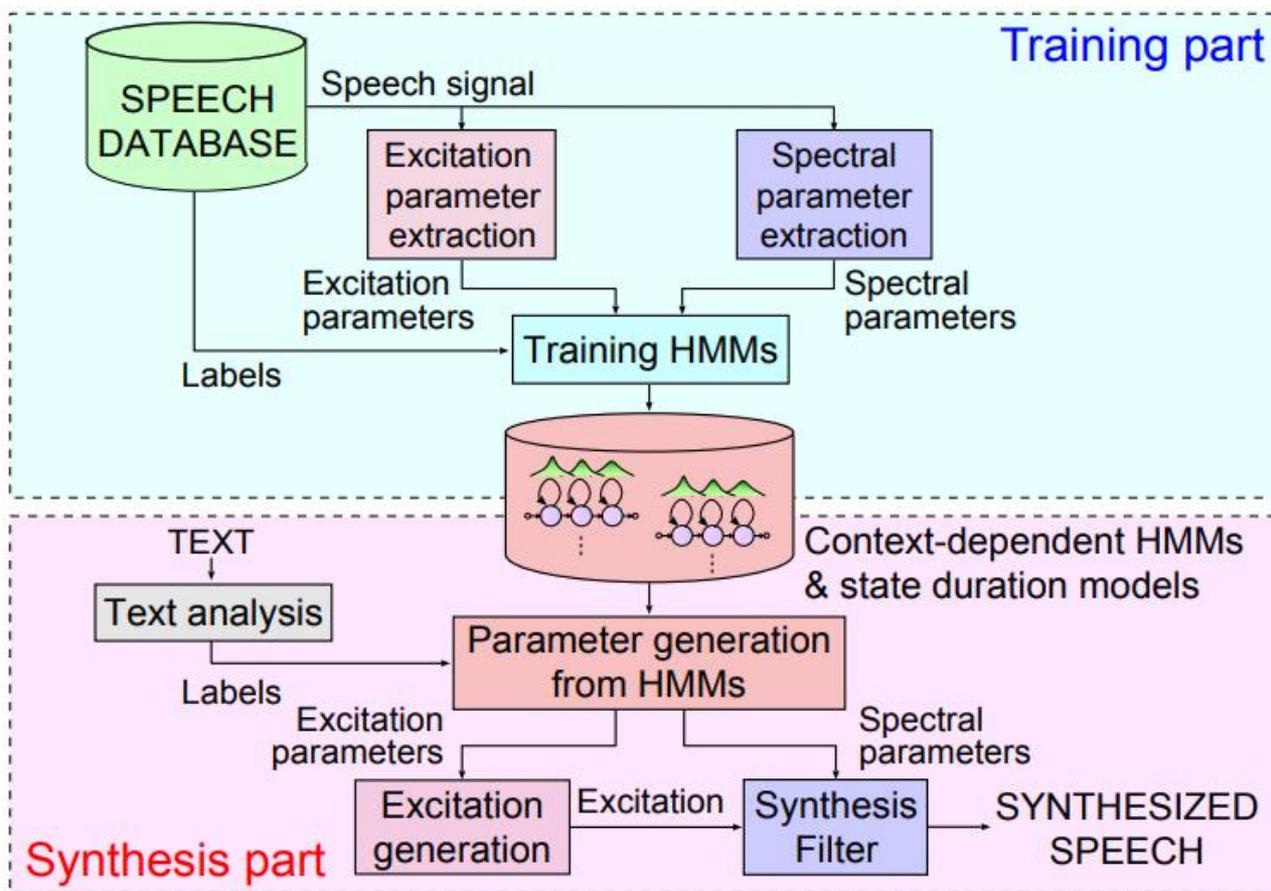


Source of image:

https://www.cs.cmu.edu/~pmuthuku/mls_p_page/lectures/spss_specom.pdf

Parametric Approach

HMM/DNN-based Speech Synthesis System (HTS)



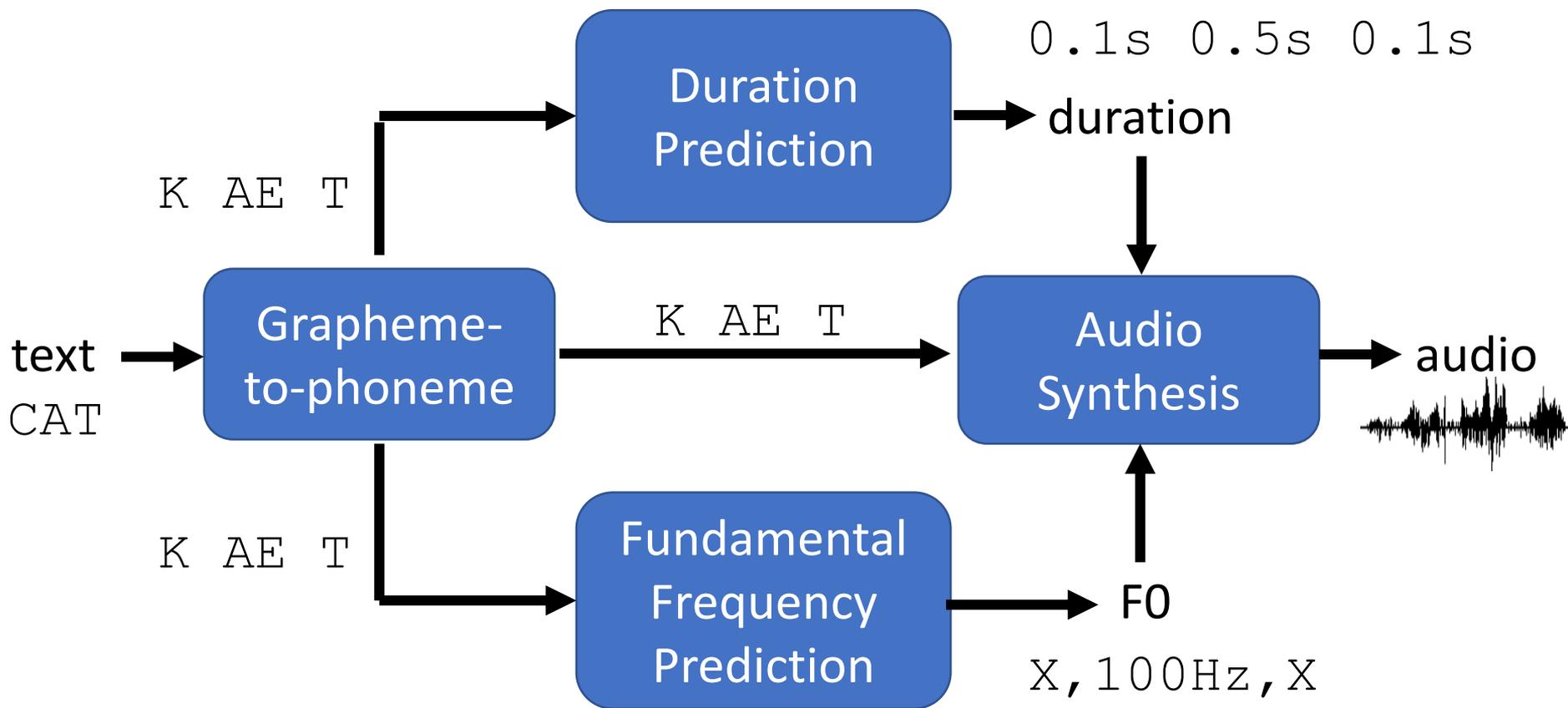
Source of image: <http://hts.sp.nitech.ac.jp/?Tutorial>

Deep Voice

[Arik, et al., ICML'17]

Deep Voice 3 is end-to-end.

[Ping, et al., ICLR'18]



All the components are deep learning based.

Outline

TTS before End-to-end

Tacotron: End-to-end TTS

Beyond Tacotron

Controllable TTS

Tacotron

[Wang, et al., INTERSPEECH'17]

[Shen, et al., ICASSP'18]

TACOTRON: TOWARDS END-TO-END SPEECH SYNTHESIS

Yuxuan Wang*, **RJ Skerry-Ryan***, **Daisy Stanton**, **Yonghui Wu**, **Ron J. Weiss[†]**, **Navdeep Jaitly**,

Zongheng Yang, **Ying Xiao***, **Zhifeng Chen**, **Samy Bengio[†]**, **Quoc Le**, **Yannis Agiomyrgiannakis**,

Rob Clark, **Rif A. Saurous***

Google, Inc.

{yxwang, rjryan, rif}@google.com

*These authors really like tacos.

[†]These authors would prefer sushi.



Before Tacotron ...

- Tacotron:
 - Input: character
 - Output: (linear) spectrogram
- First Step Towards End-to-end Parametric TTS Synthesis [\[Wang, et al., INTERSPEECH'16\]](#)
 - Input: phoneme
 - Output: acoustic features for STRAIGHT (vocoder)
- Char2wav [\[Sotelo, et al., ICLR workshop'17\]](#)
 - Input: character
 - Output: acoustic features for SampleRNN (vocoder)

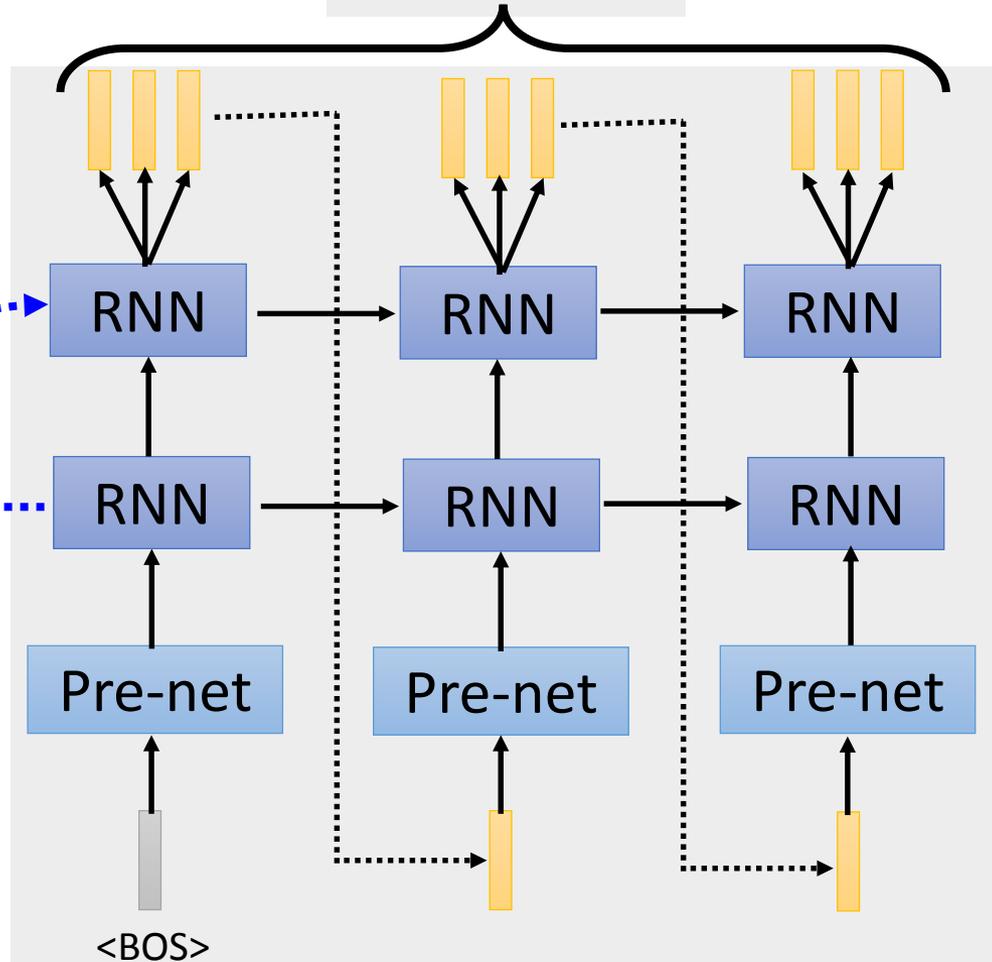
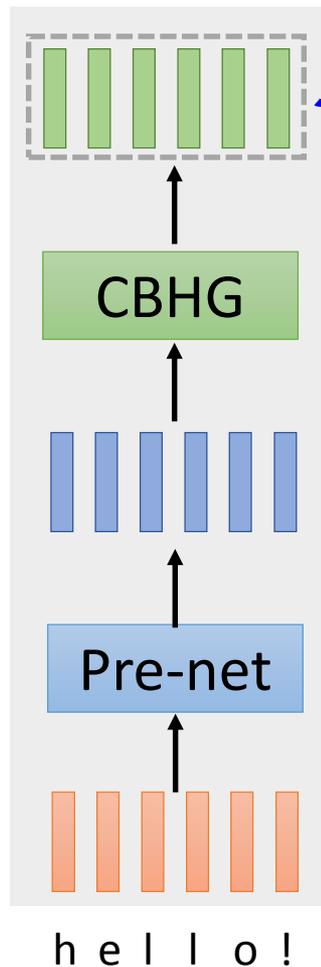
Tacotron



Post-processing

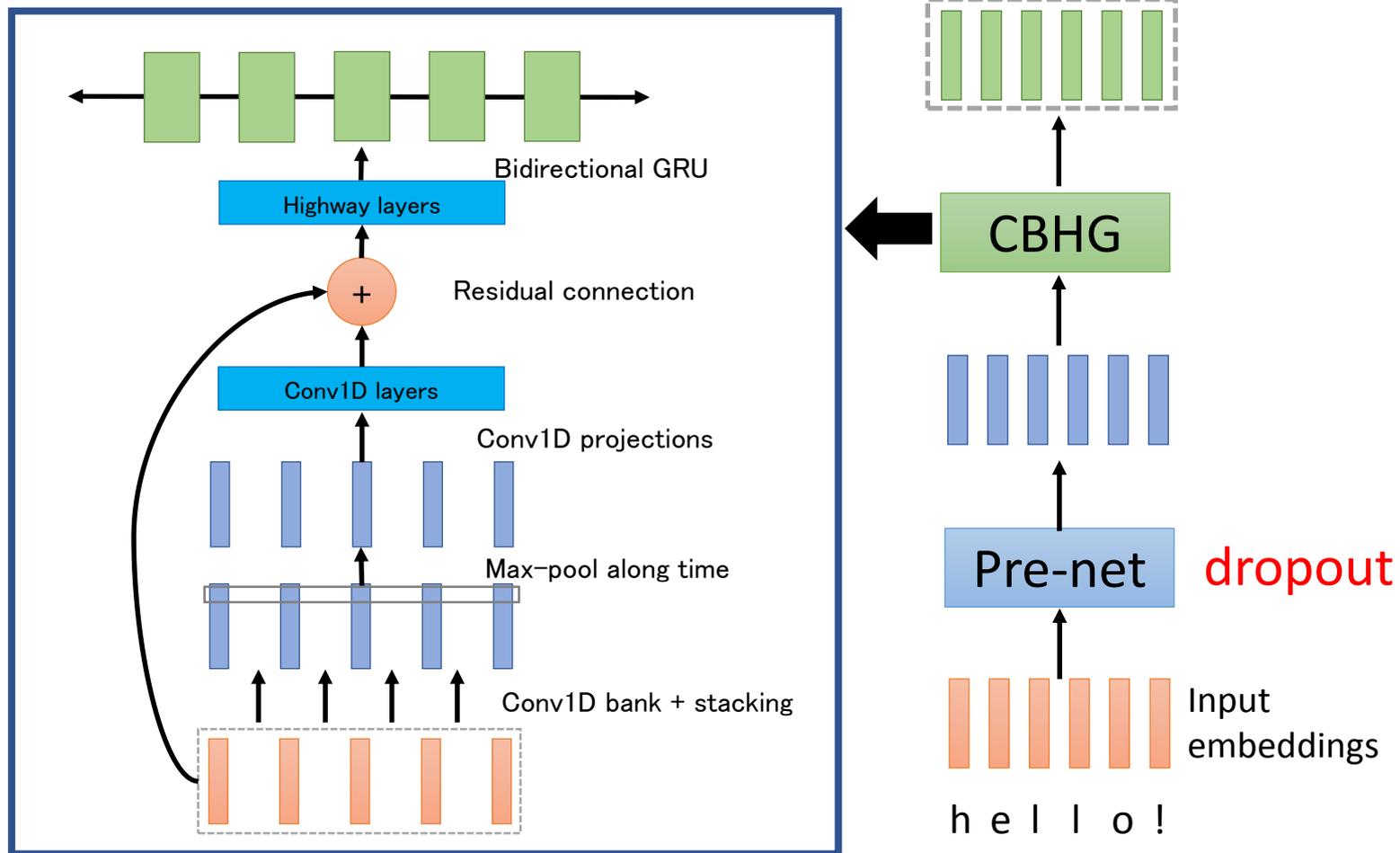


Encoder



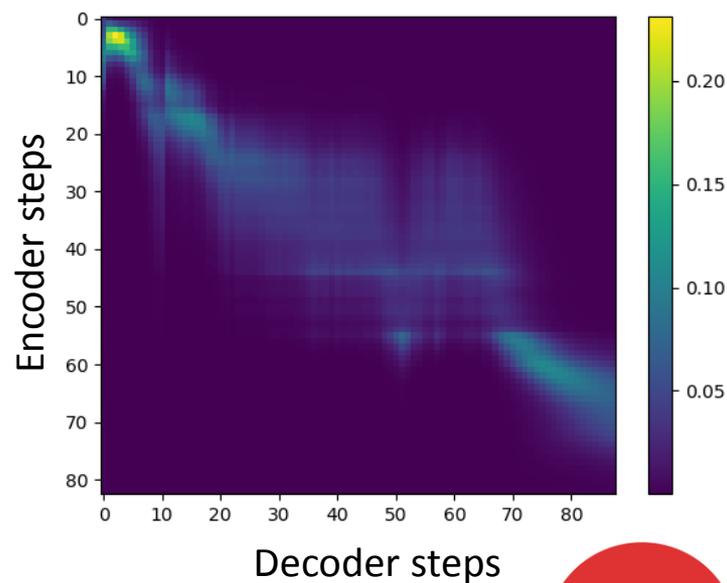
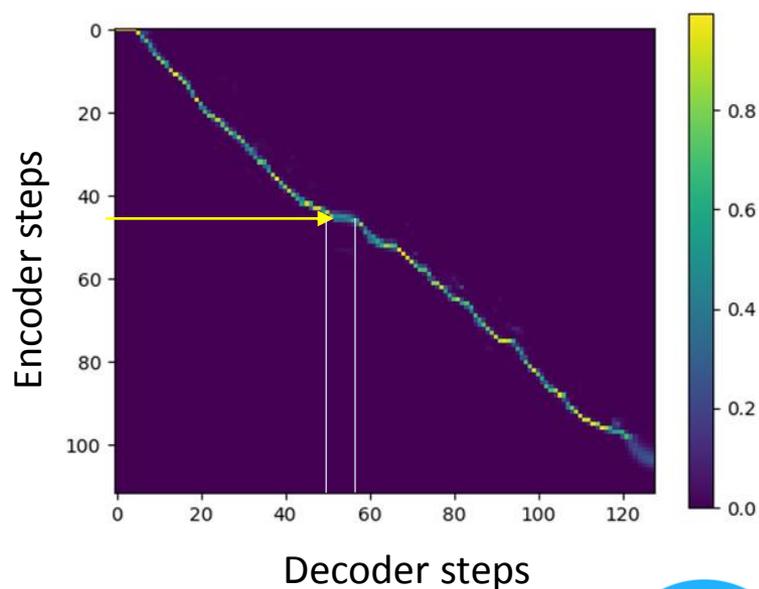
Decoder

Encoder = Grapheme-to-phoneme?

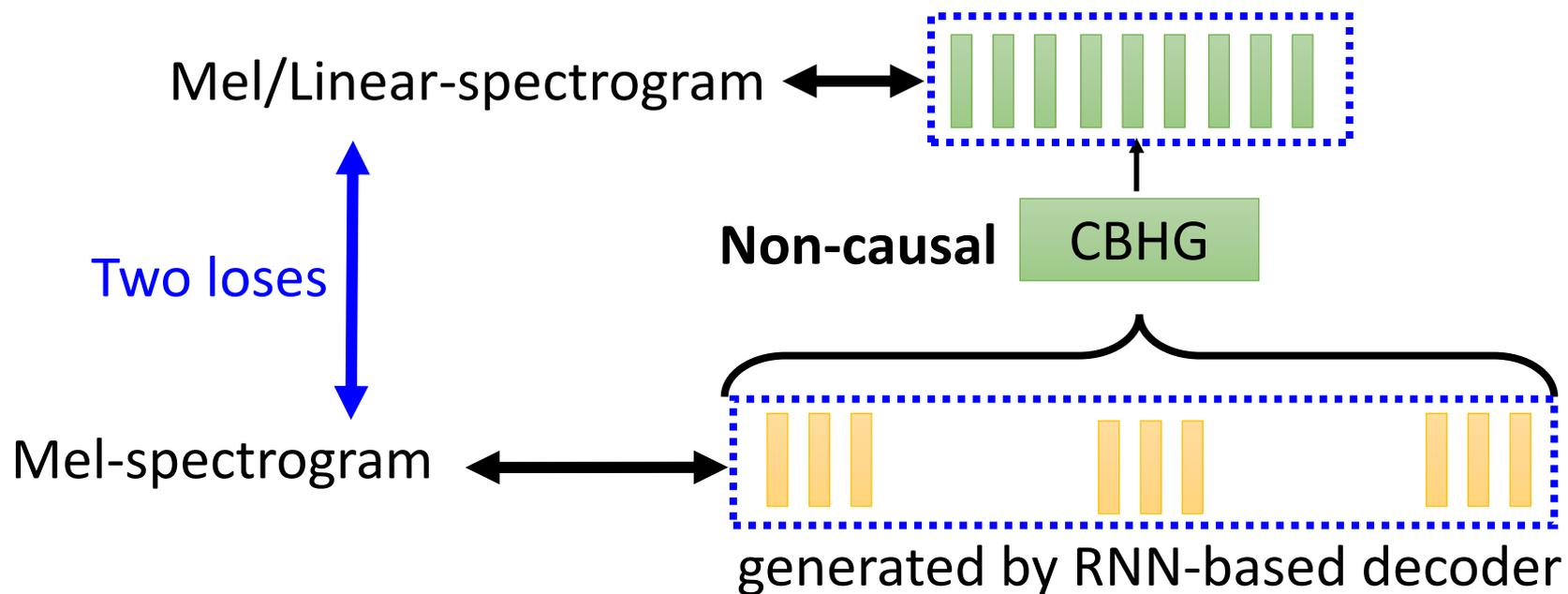


Attention = Modeling Duration ?

- The output audio and input text much be monotonic aligned.



Post processing



Vocoder:

Griffin-Lim in v1
Wavnet in v2



How good is Tacotron?

Version 1
[Wang, et al.,
INTERSPEECH'17]

	mean opinion score
Tacotron	3.82 ± 0.085
Parametric	3.69 ± 0.109
Concatenative	4.09 ± 0.119

Version 2
[Shen, et al., ICASSP'18]

System	MOS
Parametric	3.492 ± 0.096
Tacotron (Griffin-Lim)	4.001 ± 0.087
Concatenative	4.166 ± 0.091
WaveNet (Linguistic)	4.341 ± 0.051
Ground truth	4.582 ± 0.053
Tacotron 2 (this paper)	4.526 ± 0.066

How good is Tacotron?

System	MOS
Tacotron 2 (Linear + G-L)	3.944 ± 0.091
Tacotron 2 (Linear + WaveNet)	4.510 ± 0.054
Tacotron 2 (Mel + WaveNet)	4.526 ± 0.066

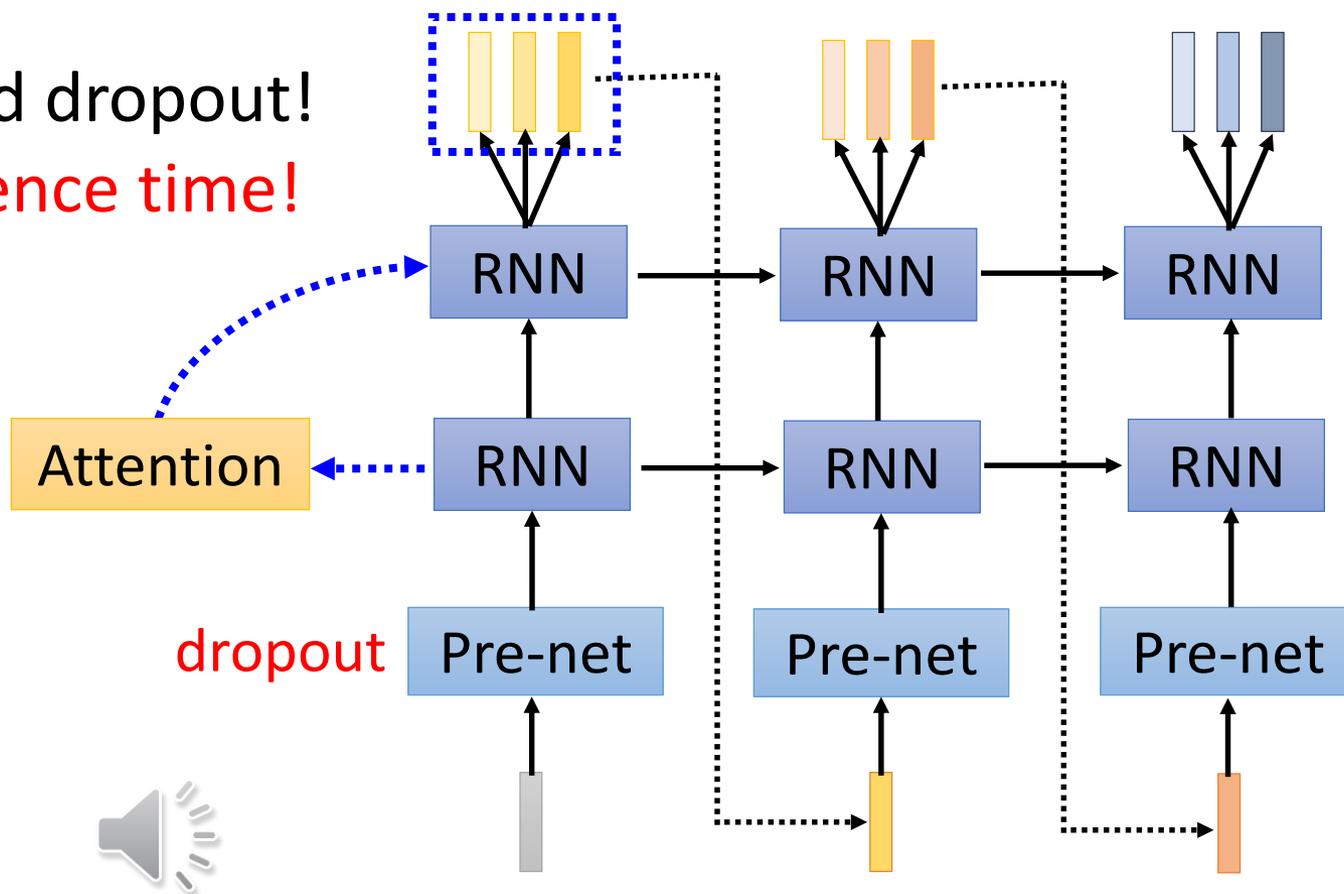
WaveNet is much better than Griffin-Lim

Training	Synthesis	
	Predicted	Ground truth
Predicted	4.526 ± 0.066	4.449 ± 0.060
Ground truth	4.362 ± 0.066	4.522 ± 0.055

WaveNet needs to be trained

Tip at Inference Phase

- You need dropout!
At inference time!



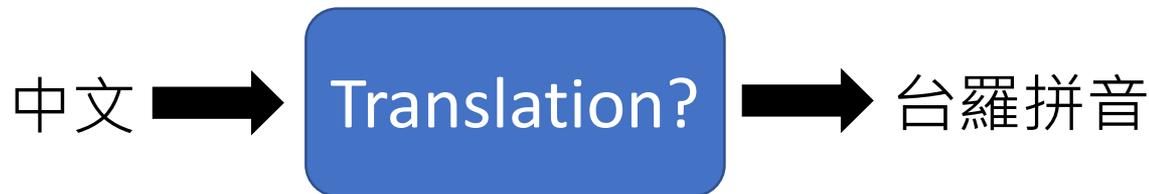
with
dropout



without
dropout

感謝杜濤同學提供實驗結果

用 Tacotron 做閩南語語音合成



https://i3thuan5.github.io/tai5-uan5_gian5-gi2_kang1-ku7/index.html

台灣語言工具



Source of training data: <https://suisiann-dataset.ithuan.tw/>

台灣嬌聲2.0



感謝張凱為同學提供實驗結果

Outline

TTS before End-to-end

Tacotron: End-to-end TTS

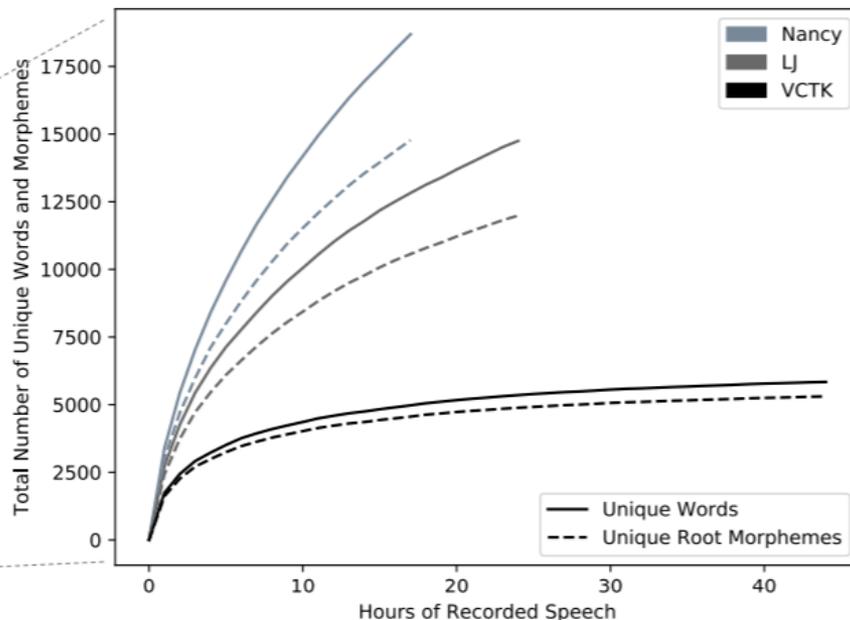
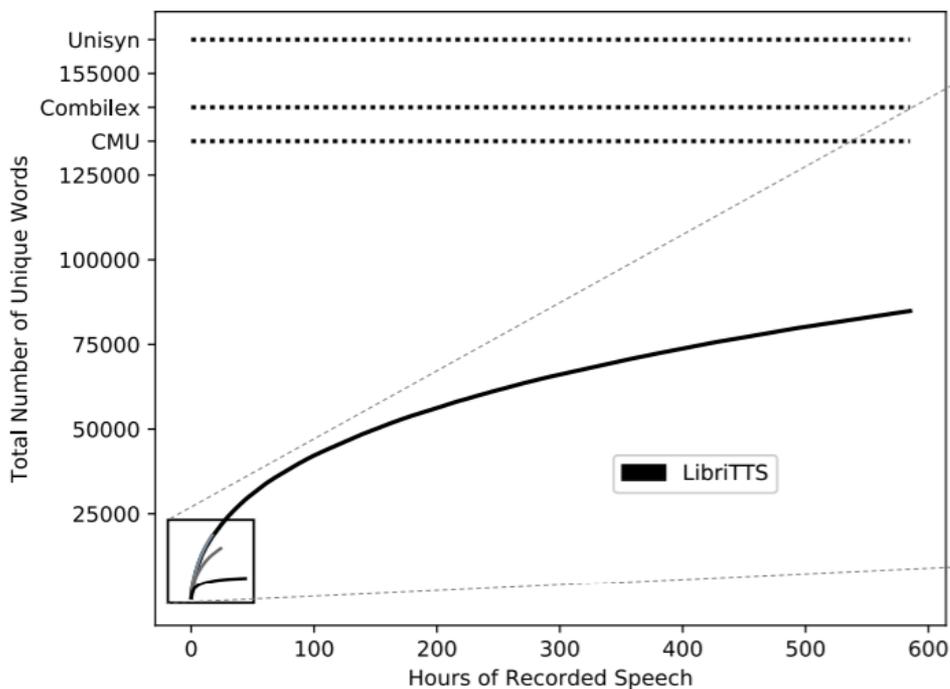
Beyond Tacotron

Controllable TTS

Mispronunciation

- The raters considered ground truth is better than Tacotron 2 because ...
- “... occasional mispronunciation by our system is the primary reason ...”

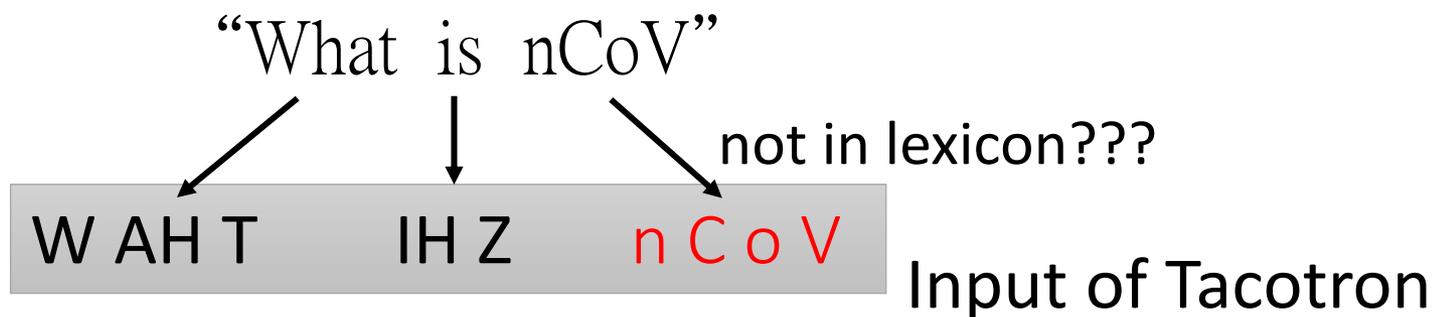
(LibriTTS dataset 585 hours)



[Taylor, et al., INTERSPEECH'19]

Mispronunciation

- Using a lexicon to transform word to phoneme, and using phoneme as Tacotron input
 - But lots of OOV words ...



- Character and phoneme hybrid input [\[Ping, et al., ICLR'18\]](#)

If the pronunciation of machine is incorrect, one can add the word into the lexicon to fix the problem.

More information for Encoder

- Syntactic information

[Guo, et al., INTERSPEECH'19]

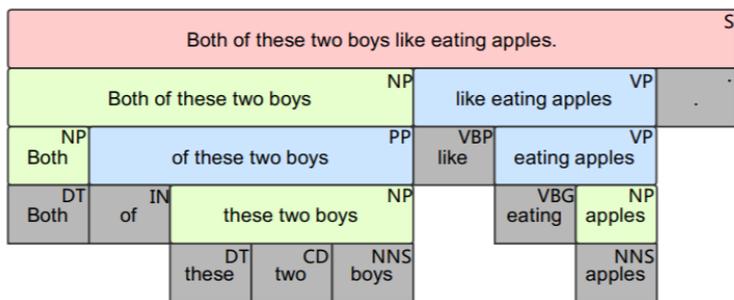


Figure 1: An example of syntactically parsed tree

一日，小龍女對楊過說：
「我也想過過過過過兒過過的生活」

Source of example:

https://youtu.be/kptTHjBi_ak

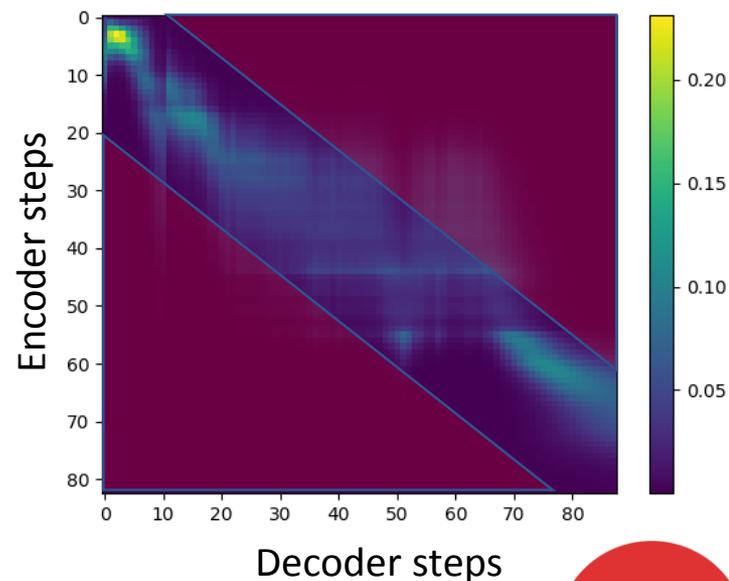
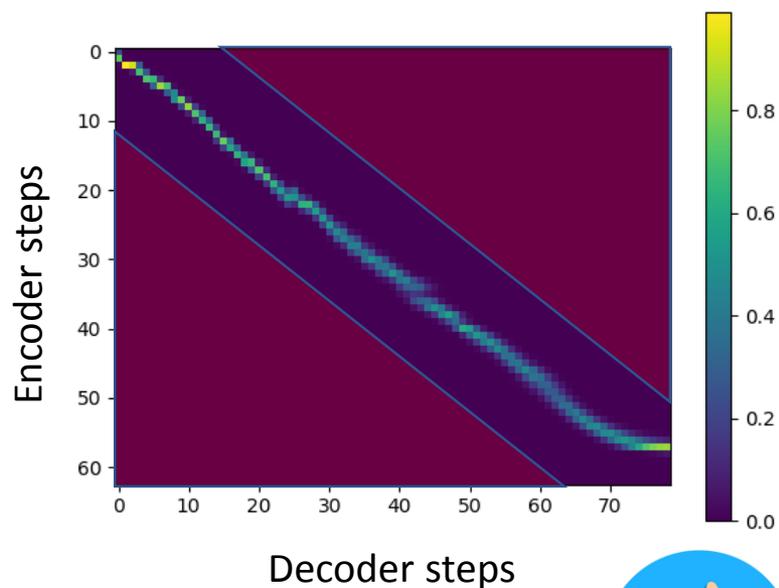
- BERT embedding as input

[Hayashi, et al., INTERSPEECH'19]

Attention

- Guided Attention [Tachibana, et al., ICASSP'18]

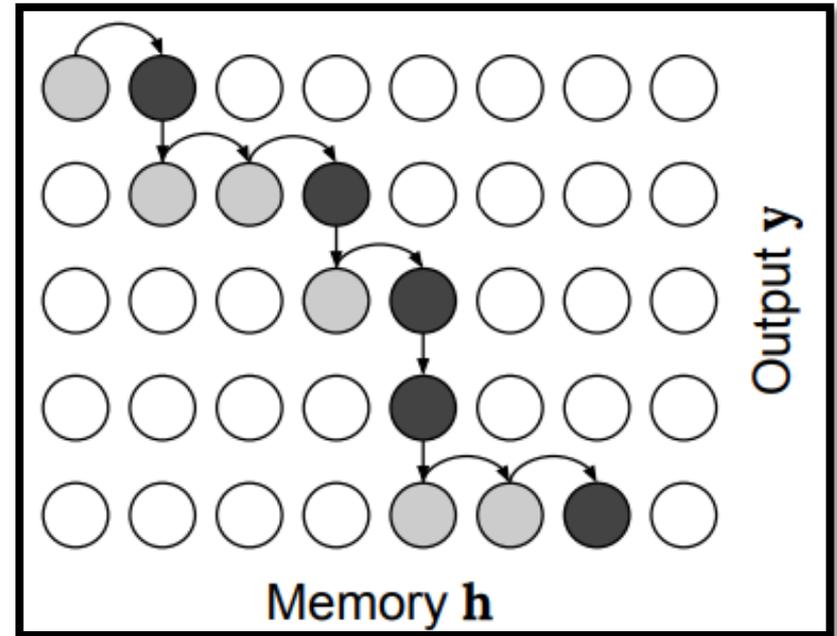
Penalizing the non-diagonal attention matrix during training



Attention

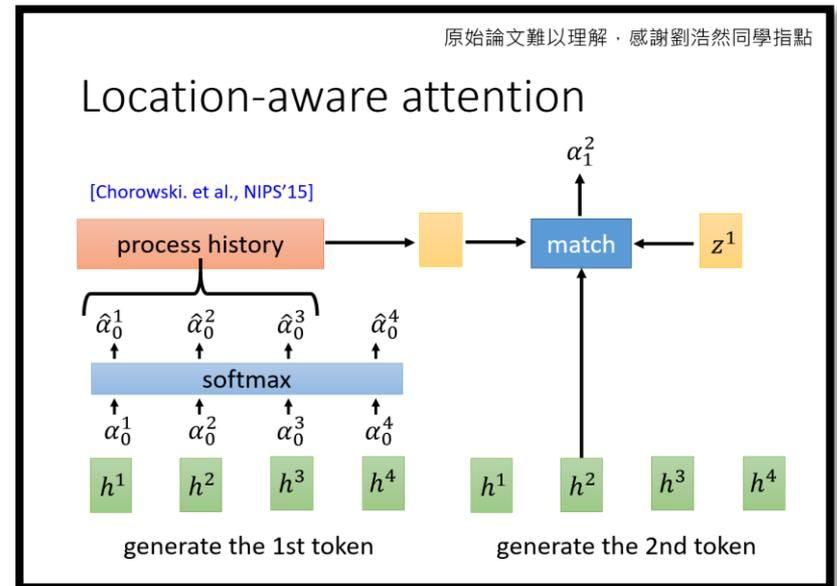
- Monotonic Attention

[Raffel, et al., ICML'17]

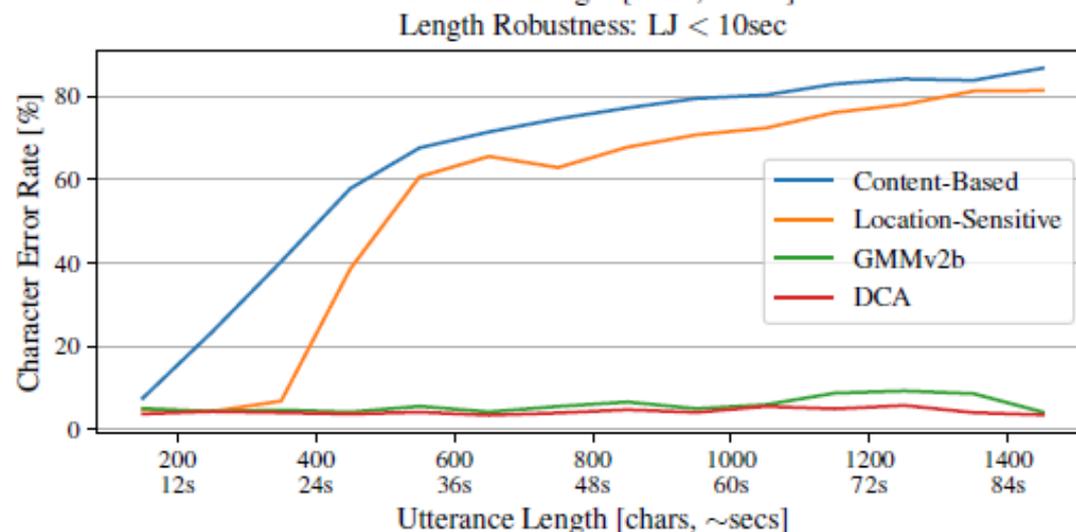
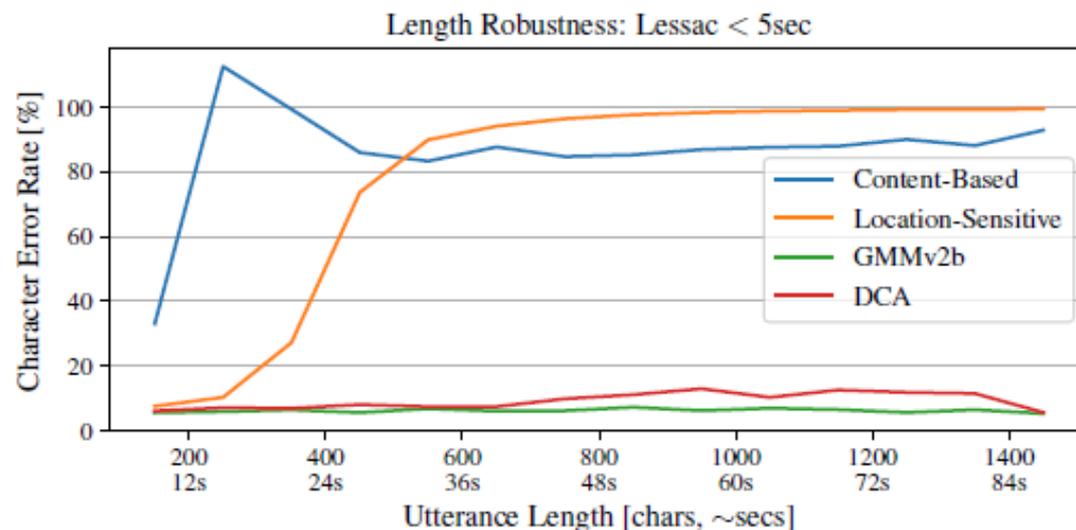


- Location-aware attention

(Have been mentioned when we talked about ASR)



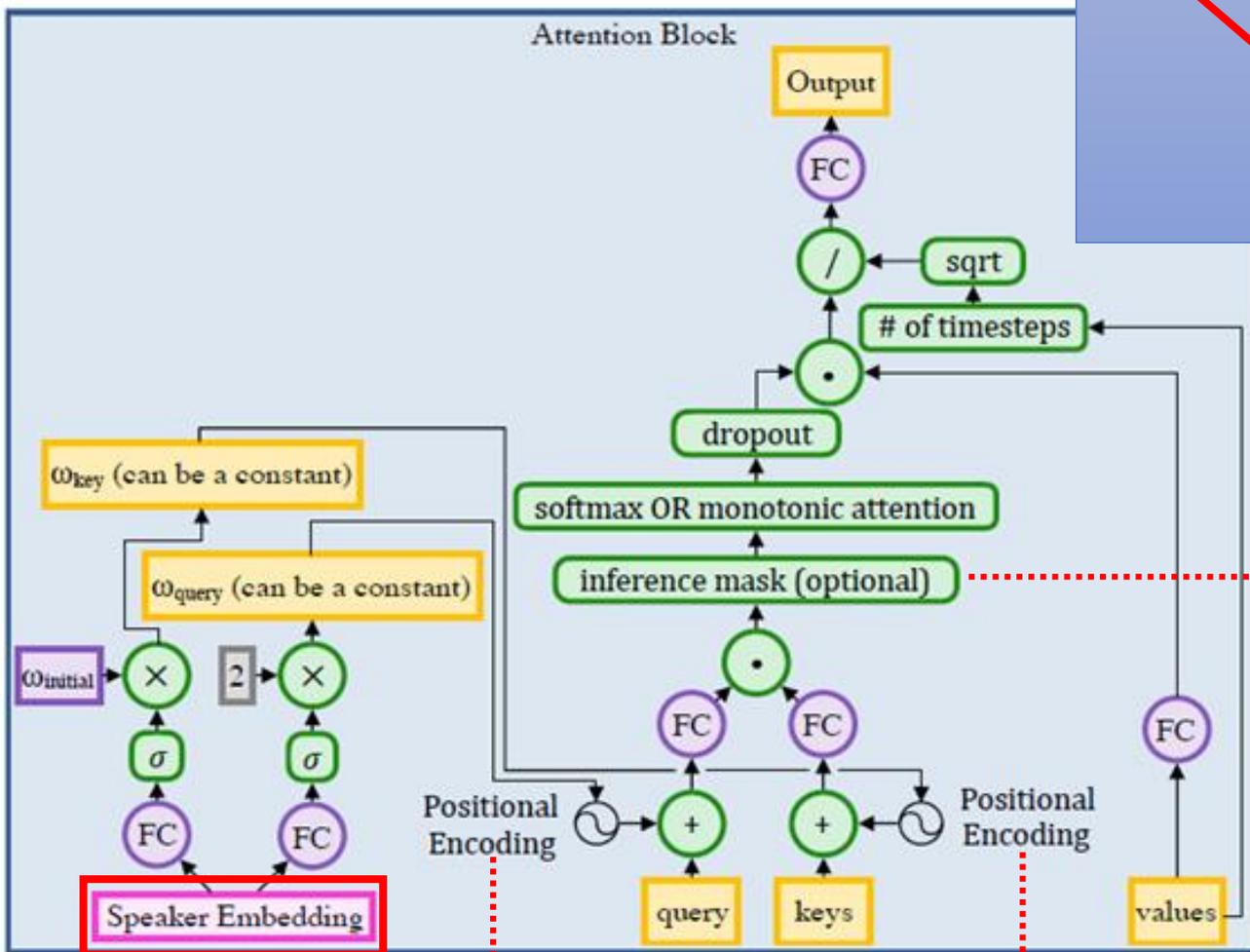
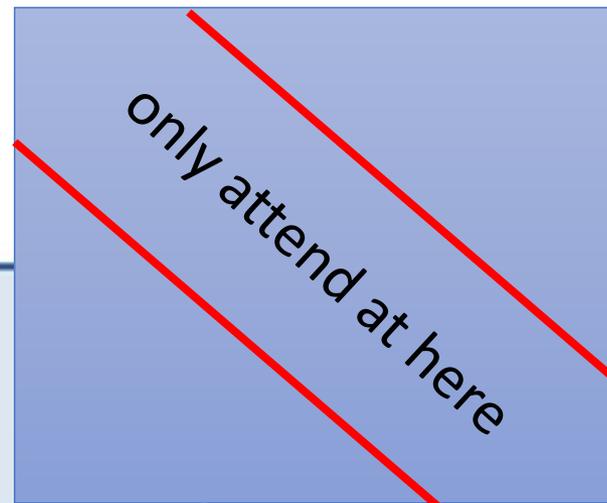
More Attention ...



[Battenberg, et al., ICASSP'20]

More ...

Attention matrix



(constraint at inference)

Only attend in a fixed window

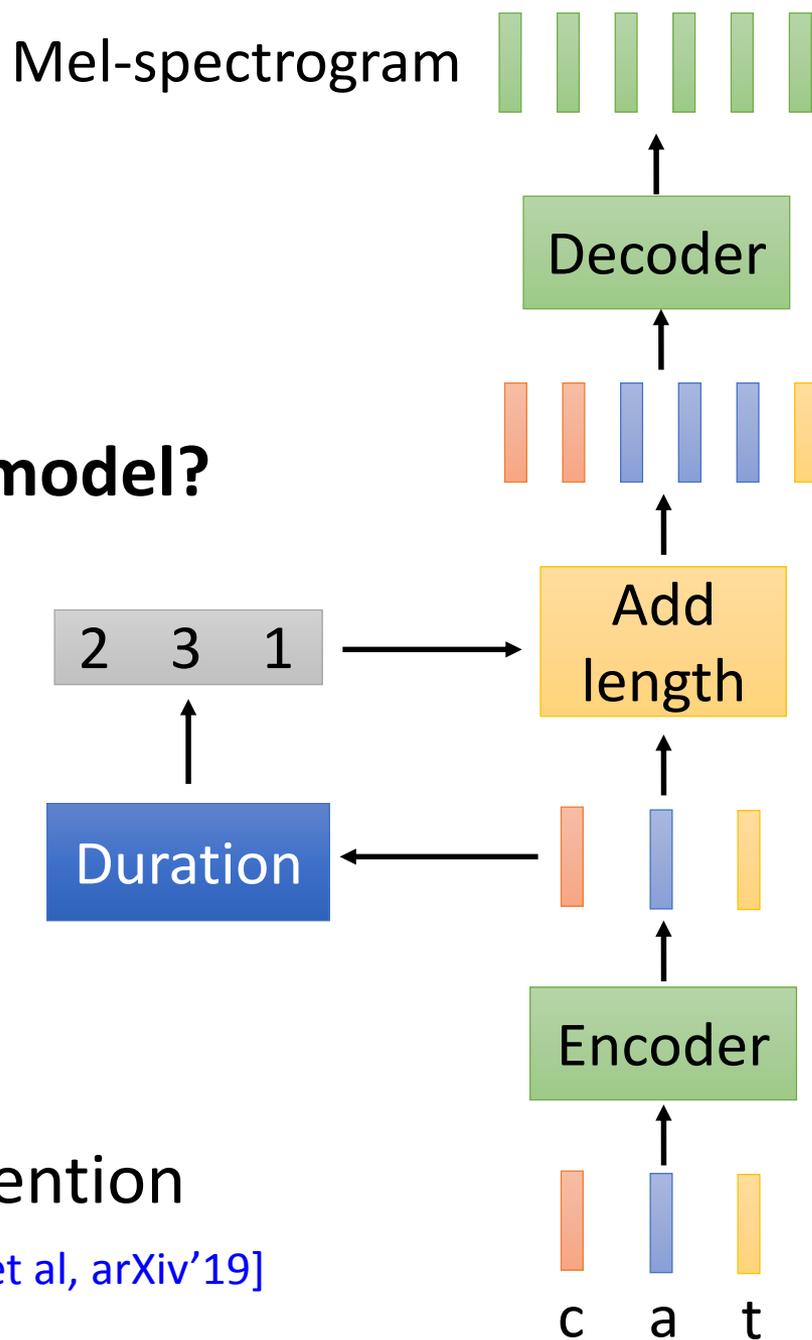
[Ping, et al., ICLR'18]

constraint attention by positional encoding

Fast Speech

[Ren, et al., NeurIPS'19]

How to train this model?



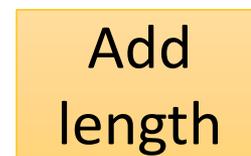
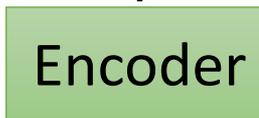
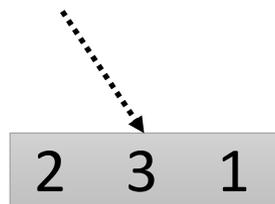
Duration Informed Attention
Network (DurlAN) [Yu, et al, arXiv'19]

Fast Speech

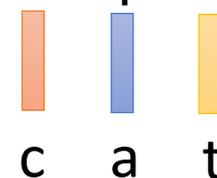
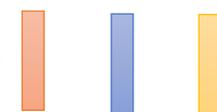
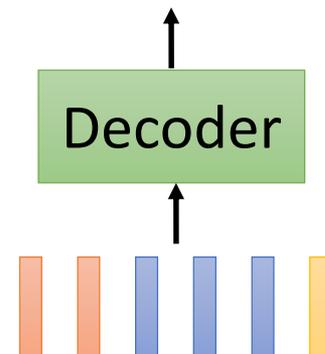
[Ren, et al., NeurIPS'19]

During the *training* phase:

Using ground truth
(alignment from
another model?)



Mel-spectrogram



Duration Informed Attention
Network (DurlAN) [Yu, et al, arXiv'19]

Fast Speech

Source of results:

<https://arxiv.org/pdf/1905.09263.pdf>

In 50 sentences:

Method	Repeats	Skips	Error Sentences	Error Rate
<i>Tacotron 2</i>	4	11	12	24%
<i>Transformer TTS</i>	7	15	17	34%
<i>FastSpeech</i>	0	0	0	0%

zero zero zero zero zero zero zero zero zero two seven nine eight F three forty zero zero zero zero zero
six four two eight zero one eight

c five eight zero three three nine a zero bf eight FALSE zero zero zero bba3add2 - c229 - 4cdb -
Calendar agent failed with error code 0x80070005 while saving appointment .

Exit process - break ld - Load module - output ud - Unload module - ignore ser - System error -
ignore ibp - Initial breakpoint -

h t t p colon slash slash teams slash sites slash T A G slash default dot aspx As always , any
feedback , comments ,

two thousand and five h t t p colon slash slash news dot com dot com slash i slash n e slash f d
slash two zero zero three slash f d

Dual Learning: ASR & TTS

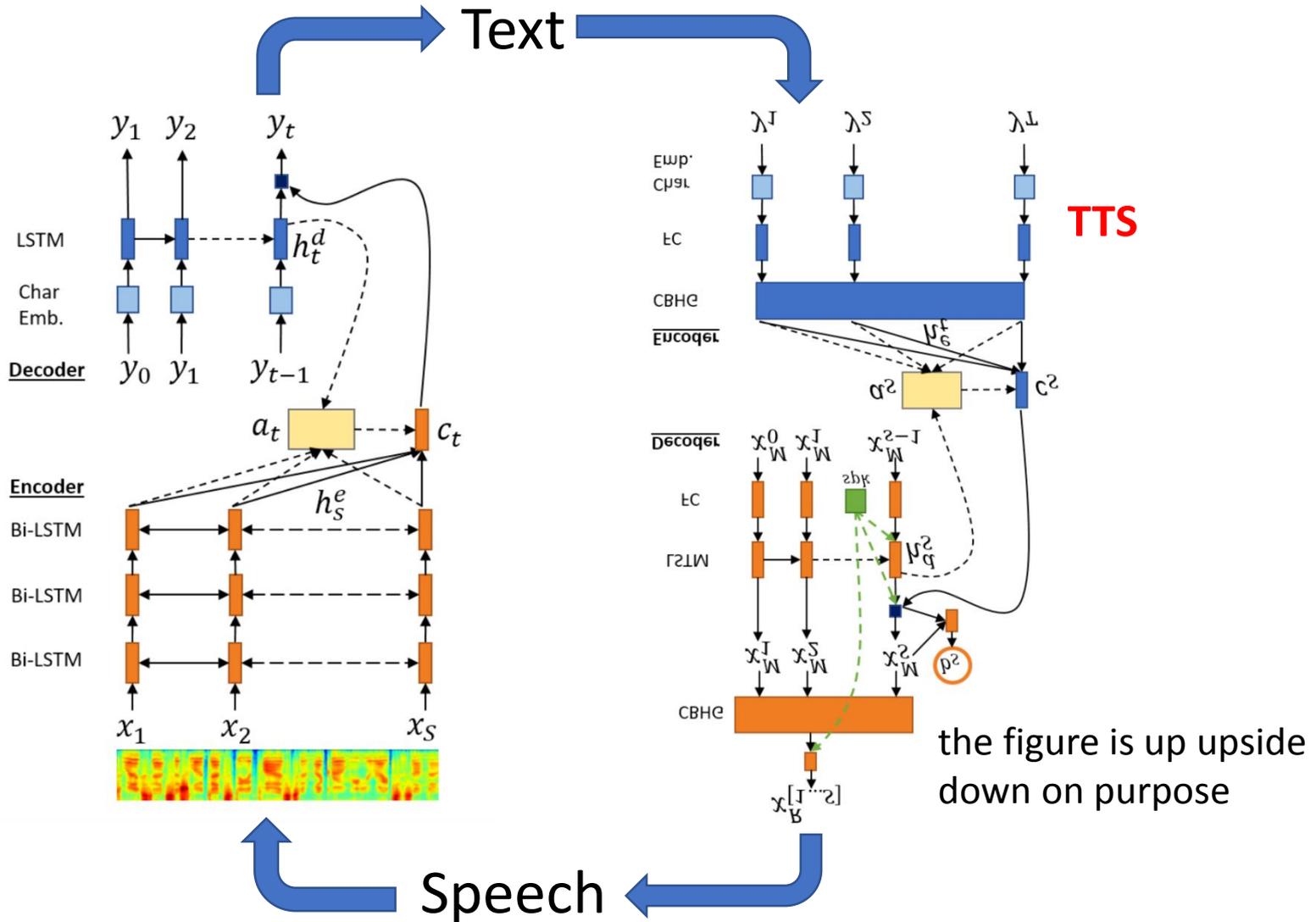
ASR & TTS form a cycle.

Speech Chain

[Tjandra et al., ASRU 2017]

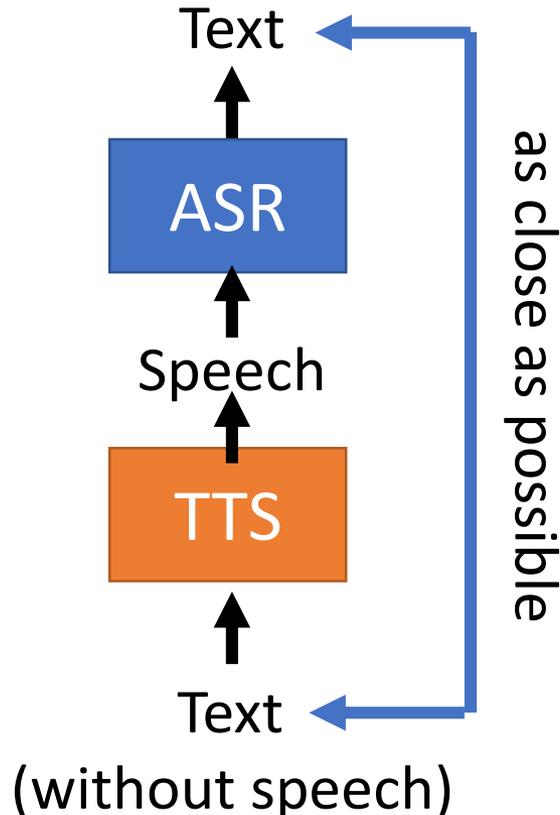
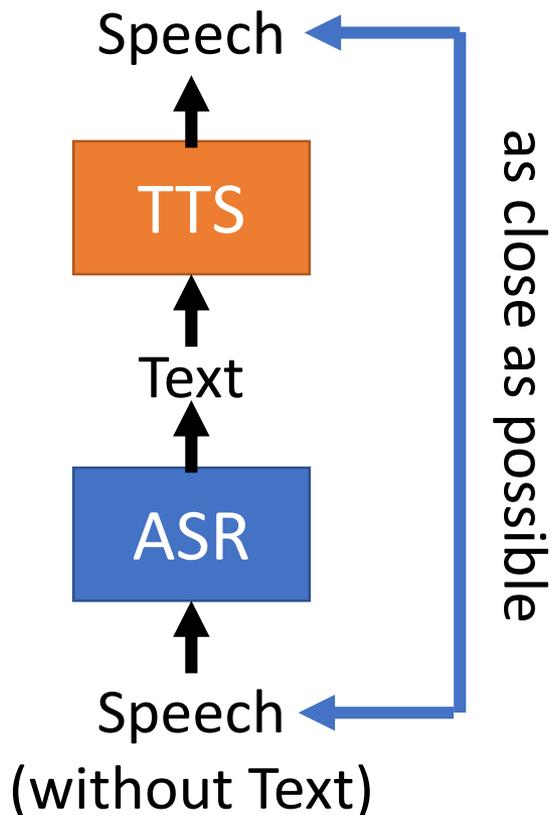
ASR

TTS



Dual Learning: TTS v.s. ASR

- Given pretrained TTS and ASR system



Dual Learning: TTS v.s. ASR

- Experiments

Mel: mel-spectrogram

Raw: raw waveform

Table 2: Experiment result for multi-speaker test set.

1600 utterance-sentence pairs

7200 unpaired utterances and sentences

Data	Hyperparameters			ASR	TTS		
	α	β	gen. mode	CER (%)	Mel	Raw	Acc (%)
Paired (80 utt/spk)	-	-	-	26.47	10.213	13.175	98.6
+ Unpaired (remaining)	0.25	1	greedy	23.03	9.137	12.863	98.7
	0.5	1	greedy	20.91	9.312	12.882	98.6
	0.25	1	beam 5	22.55	9.359	12.767	98.6
	0.5	1	beam 5	19.99	9.198	12.839	98.6

supervised loss

unsupervised loss

Prediction of the "end-of-utterance"

mse

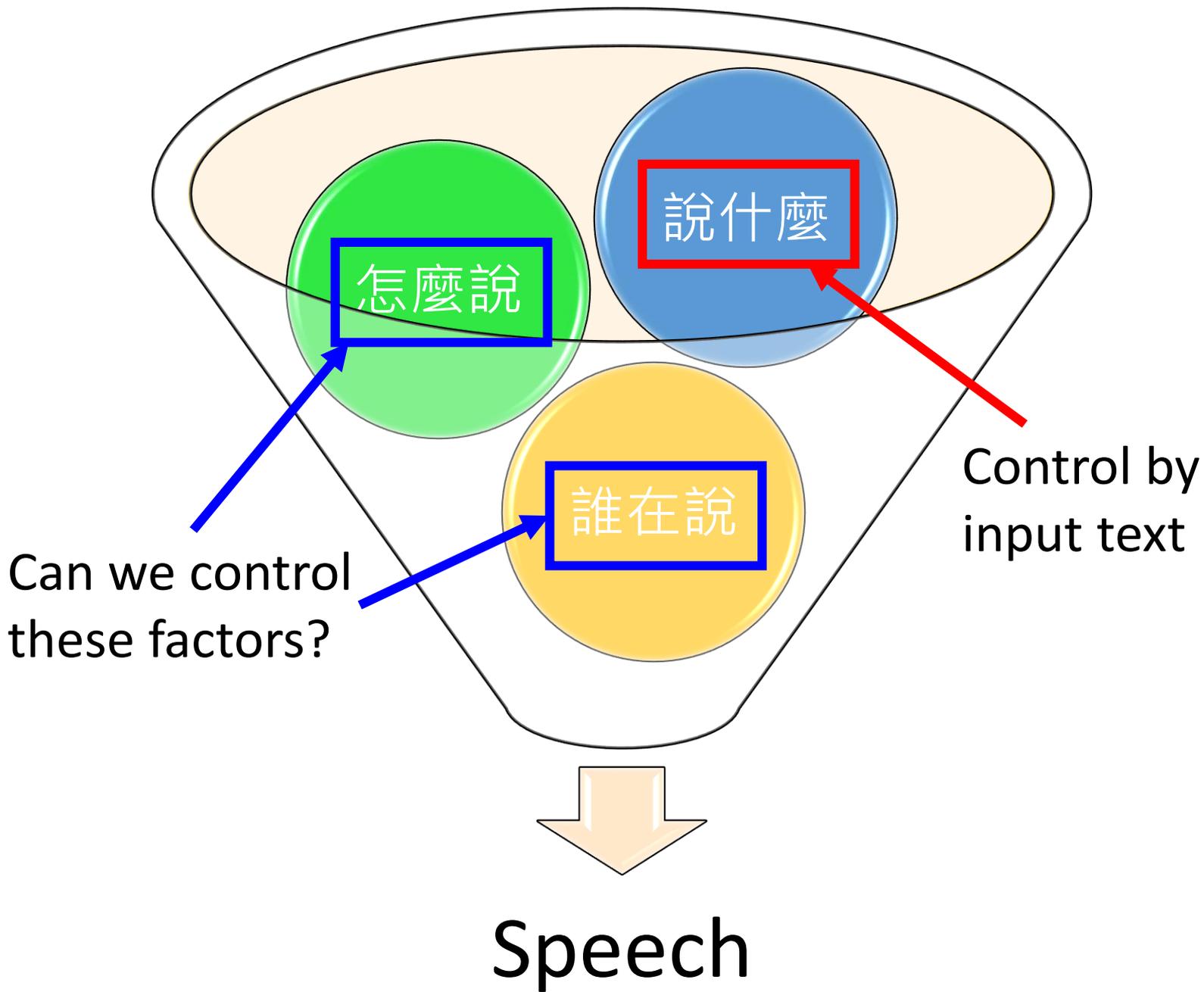
Outline

TTS before End-to-end

Tacotron: End-to-end TTS

Beyond Tacotron

Controllable TTS



Controllable TTS

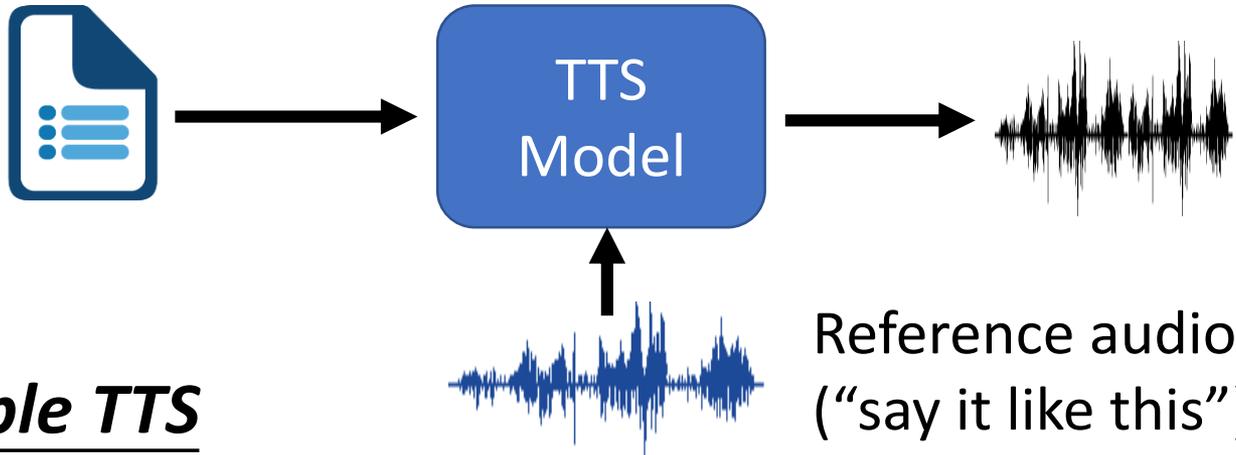
- 誰在說?
 - Synthesize speech for a specific person (voice cloning)
 - Lack of high quality single speaker data to train a speech synthesis system
- 怎麼說?
 - Intonation (語調), stress (重音), rhythm (韻律) ...
 - Prosody (抑揚頓挫)

How to describe prosody?

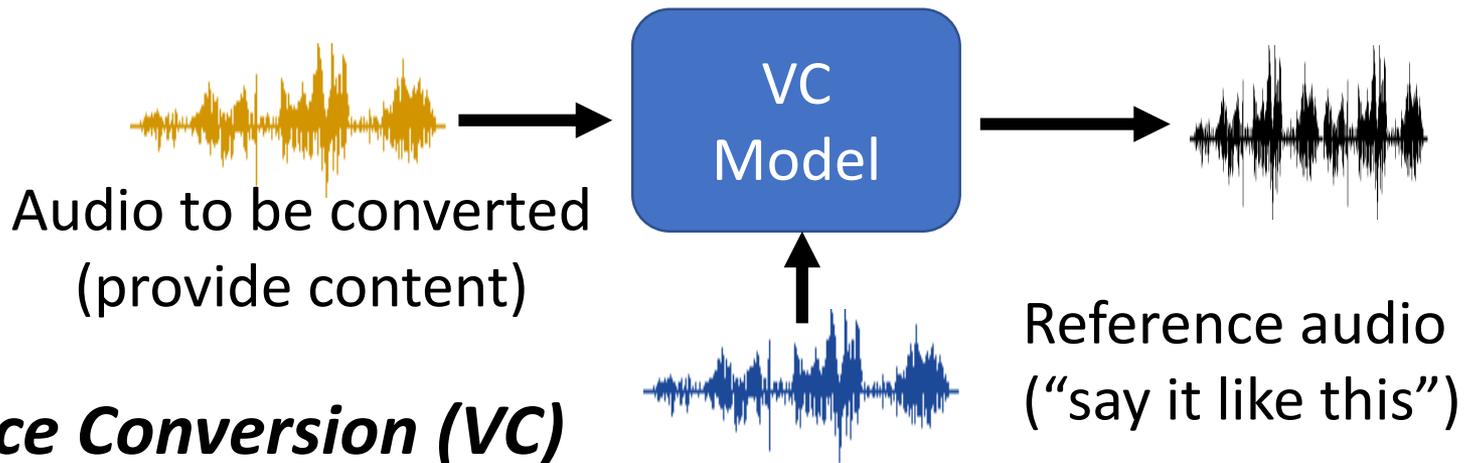
Definition. *Prosody is the variation in speech signals that remains after accounting for variation due to phonetics, speaker identity, and channel effects (i.e. the recording environment).*

[Skerry-Ryan, et al., ICML'18]

Controllable TTS

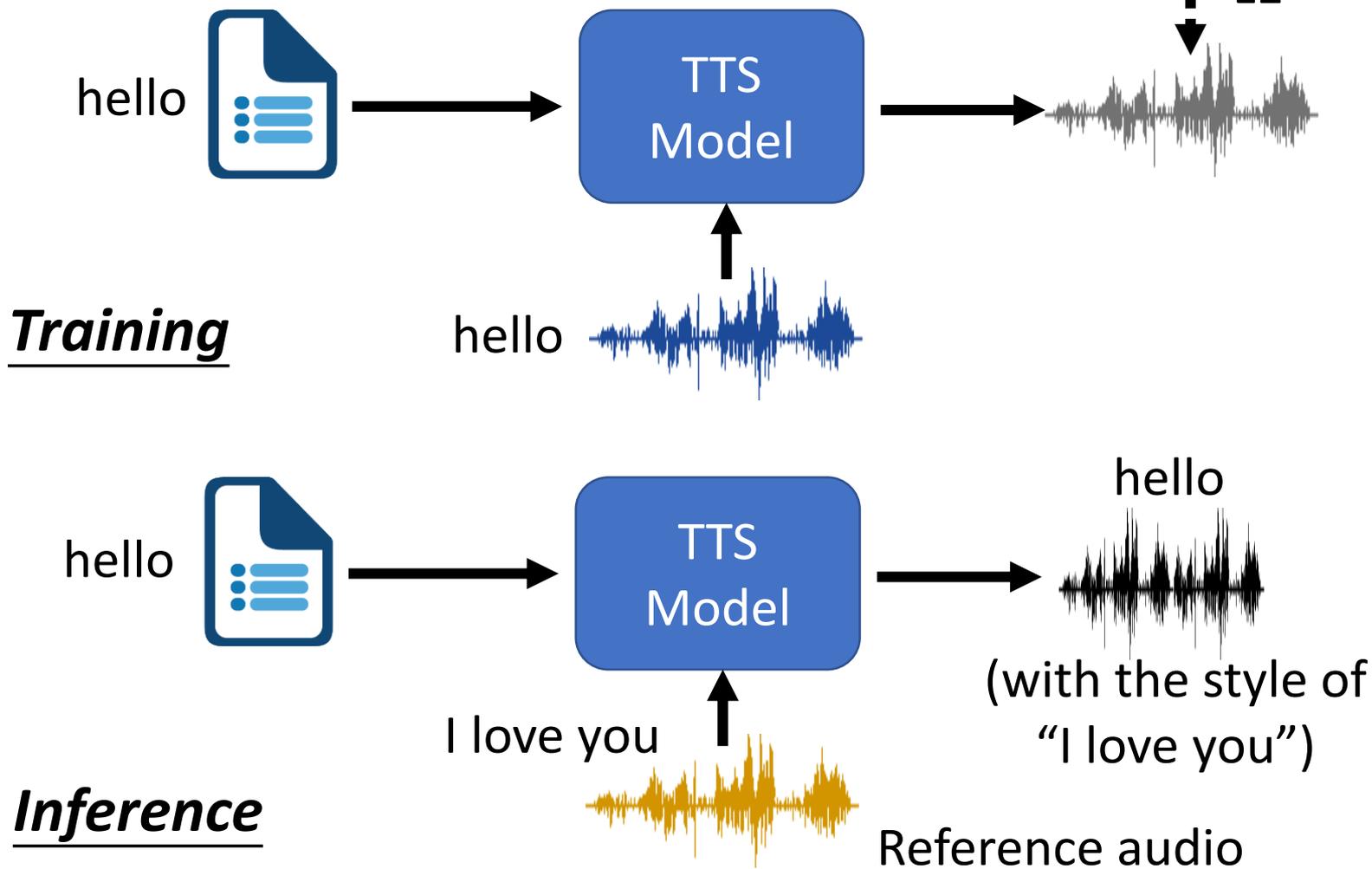


Controllable TTS

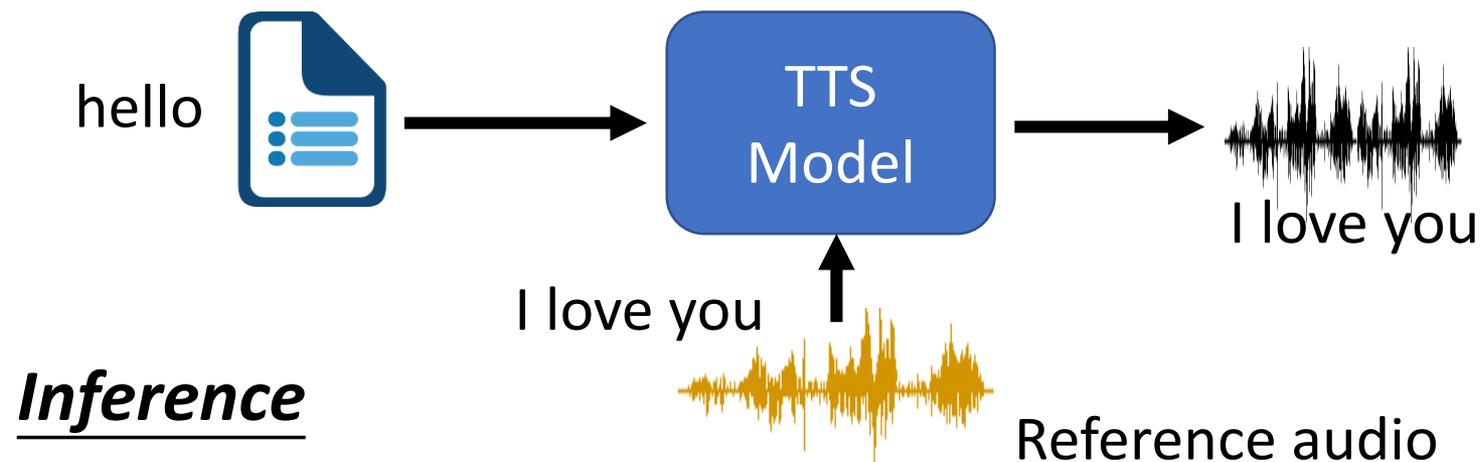
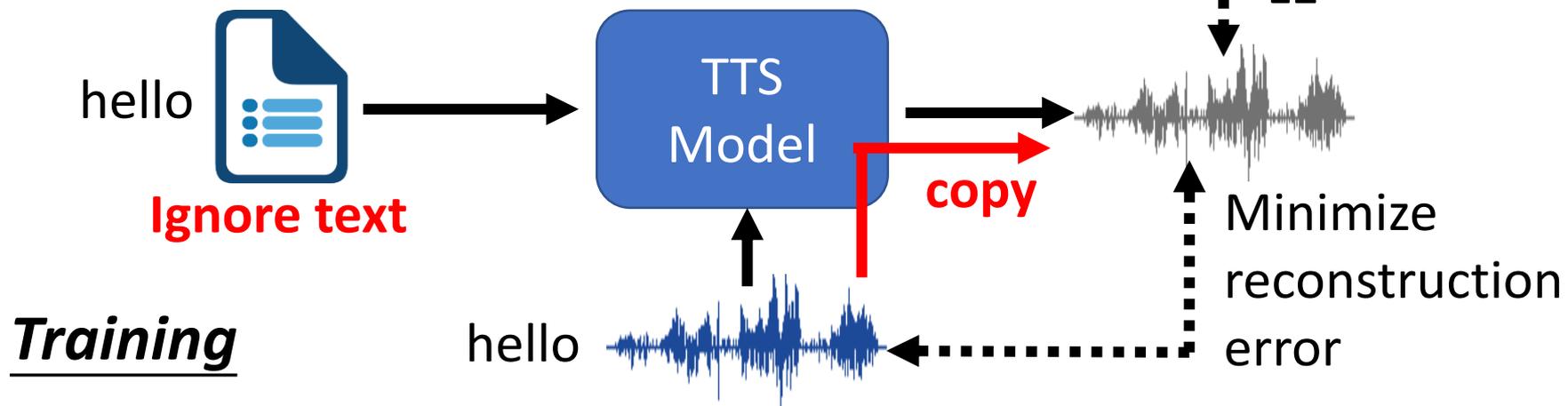


Voice Conversion (VC)

Controllable TTS

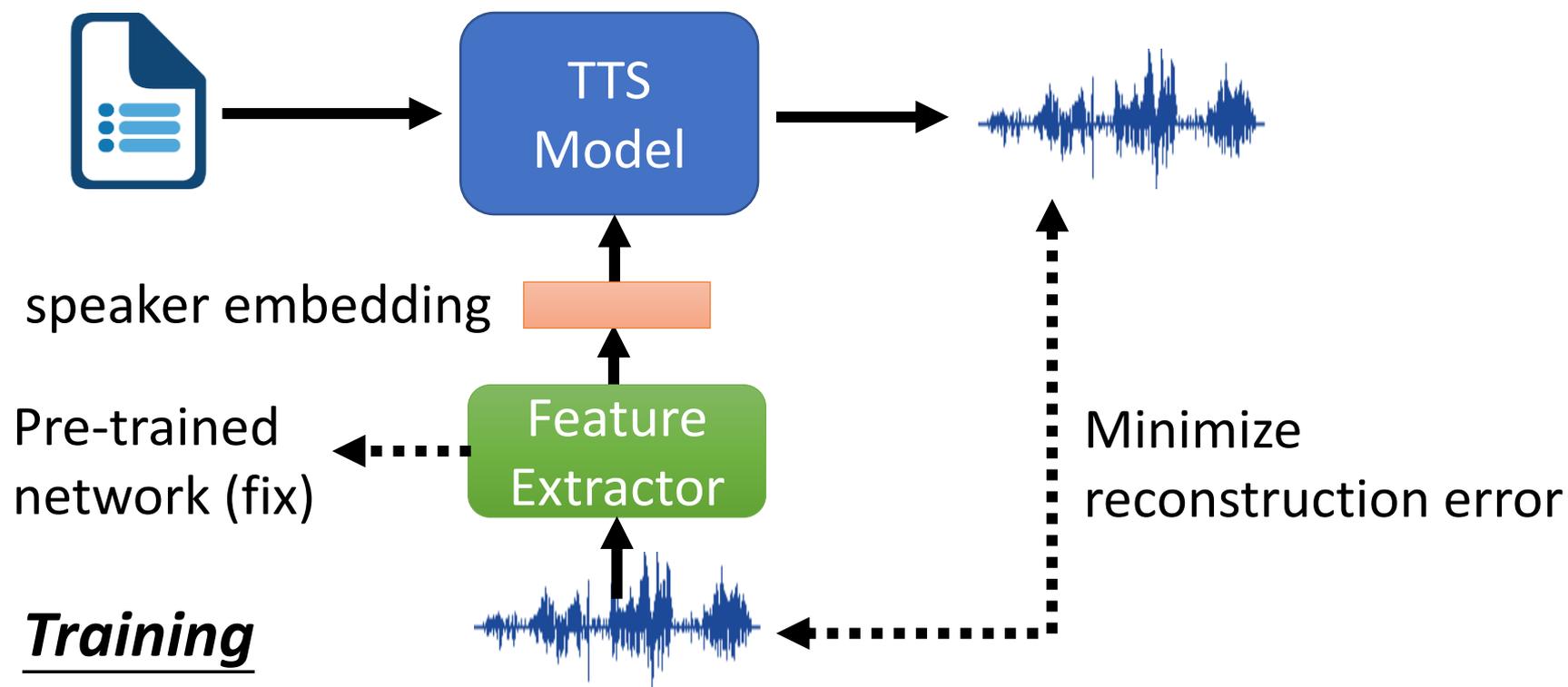


Controllable TTS



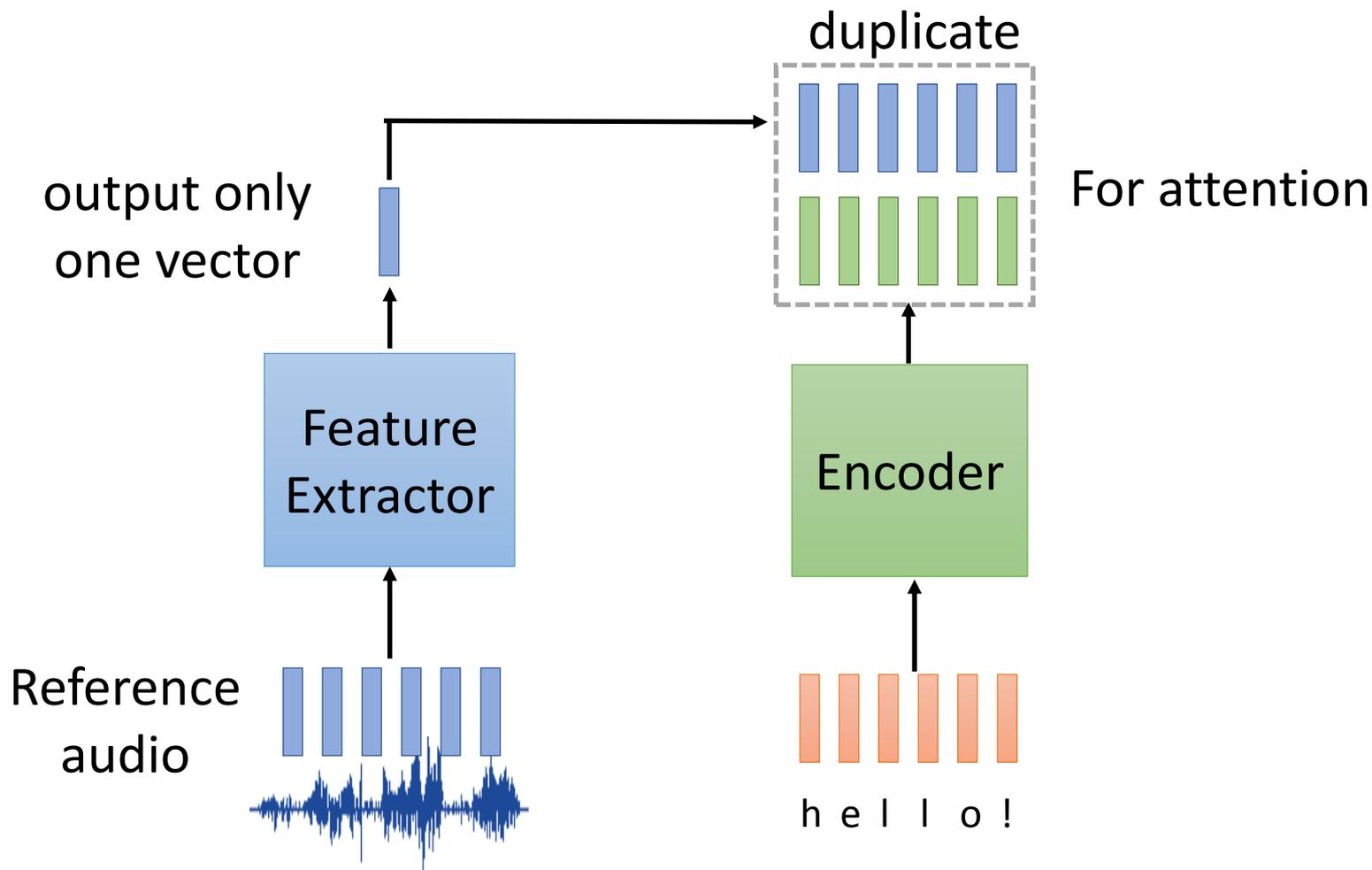
Speaker Embedding

[Jia, et al., NeurIPS'18]

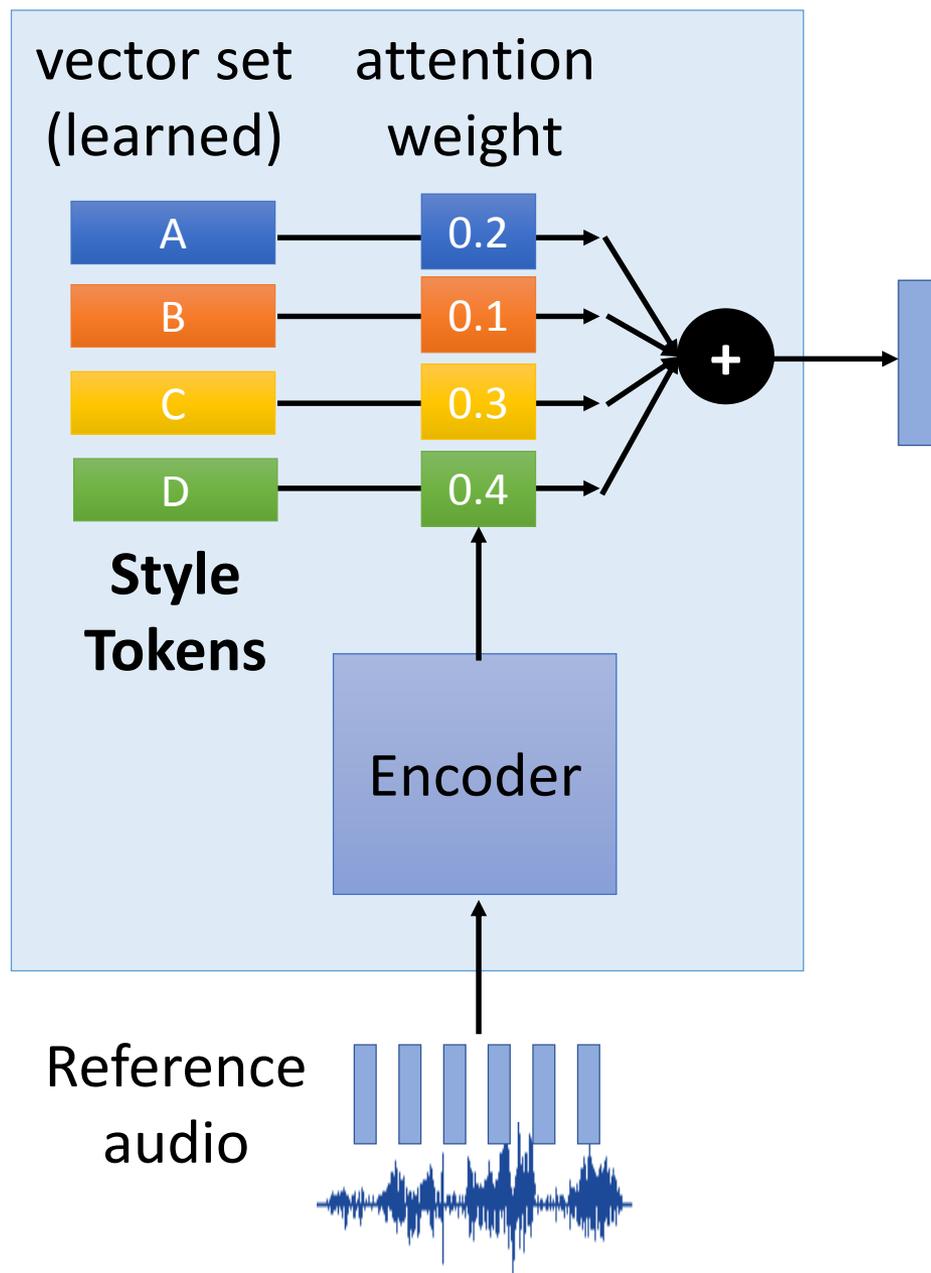
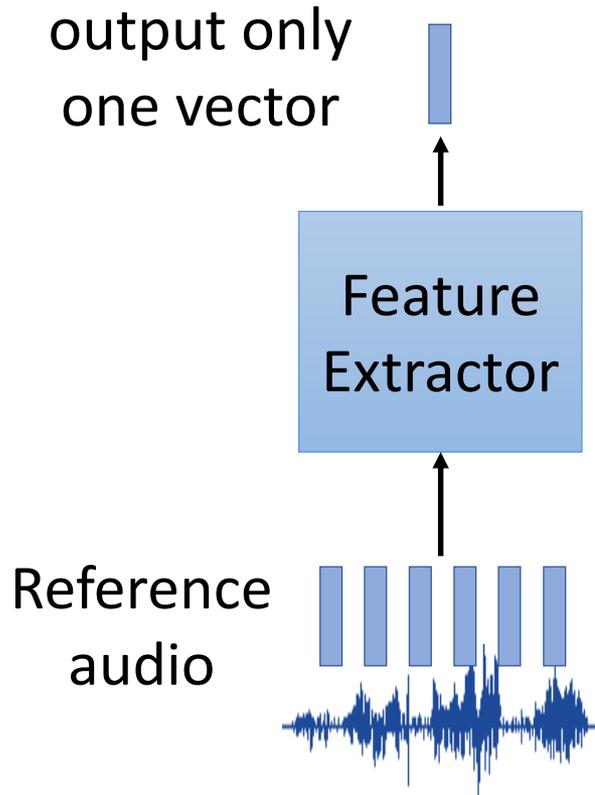


GST-Tacotron

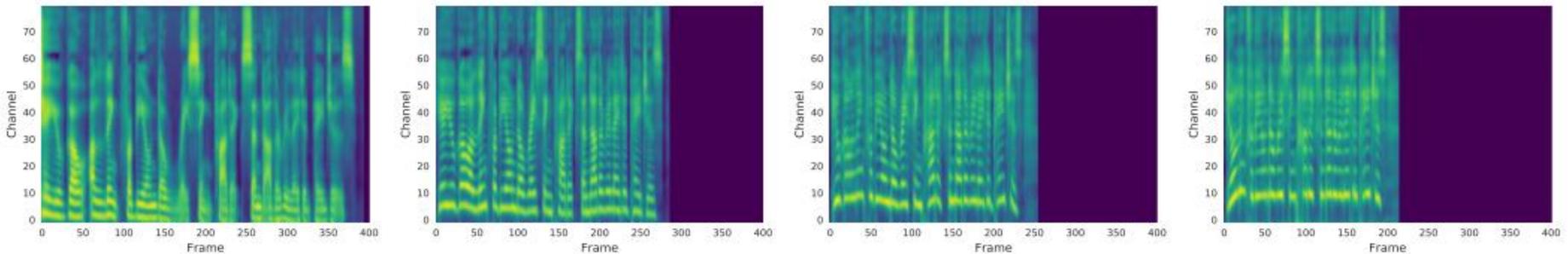
GST = global style tokens
[Wang, et al., ICML'18]



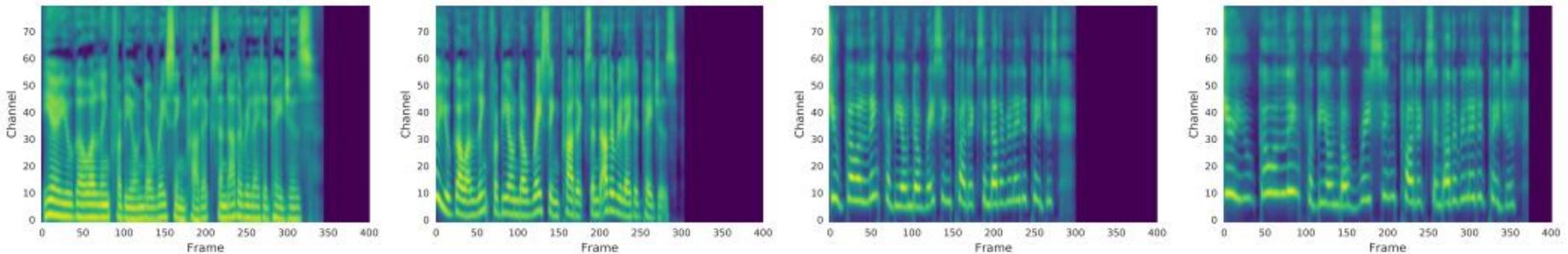
GST-Tacotron



- What does the tokens effect?
 - One token corresponds to a lower pitch voice
 - One token for a decreasing pitch
 - One token for a faster speaking rate
 -



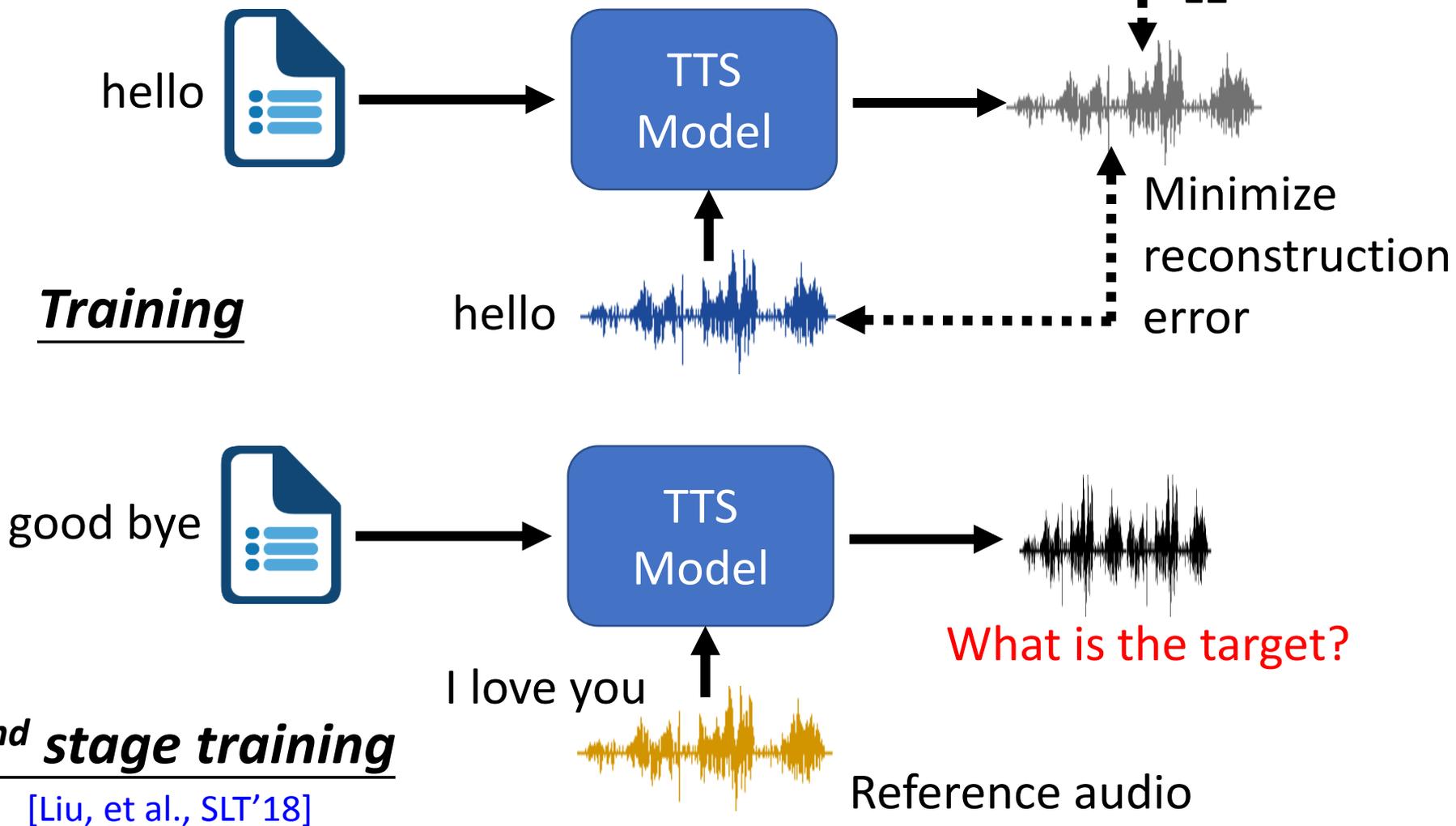
(a) Token A (speed)



(b) Token B (animated)

Source of image: <https://arxiv.org/pdf/1803.09017.pdf>

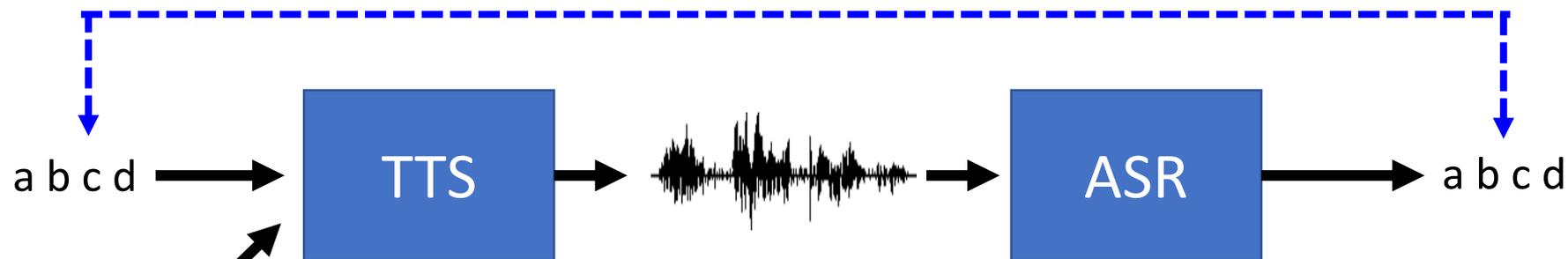
Two-stage Training



[Liu, et al., SLT'18]

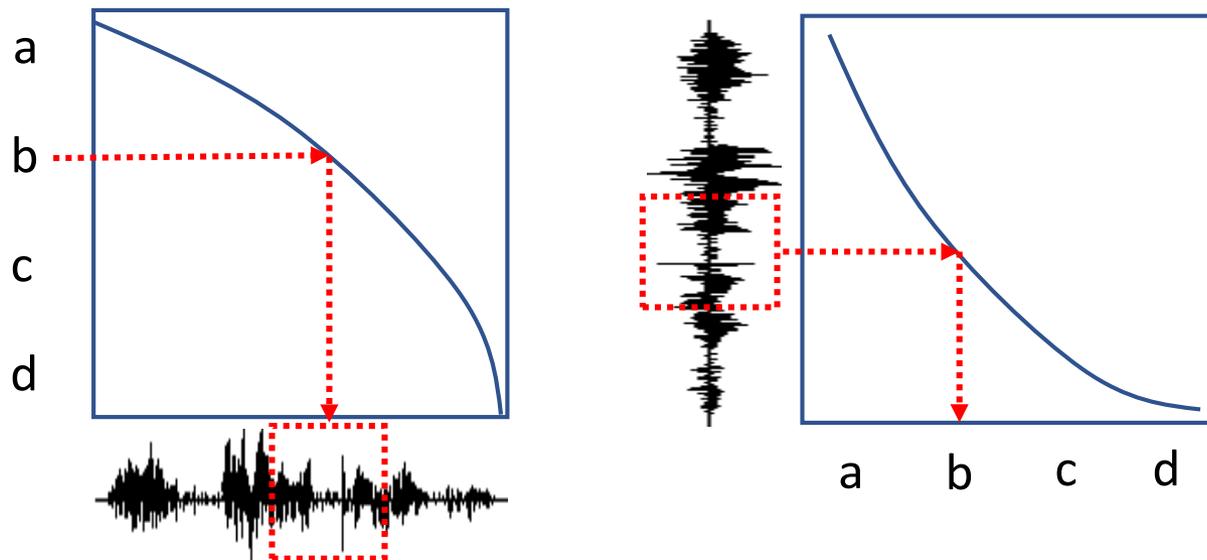
Two-stage Training

minimize recognition error



reference
audio

attention consistency



Concluding Remarks

TTS before End-to-end

Tacotron: End-to-end TTS

Beyond Tacotron

Controllable TTS

Reference

- [Wang, et al., INTERSPEECH'17] Yuxuan Wang, R.J. Skerry-Ryan, Daisy Stanton, Yonghui Wu, Ron J. Weiss, Navdeep Jaitly, Zongheng Yang, Ying Xiao, Zhifeng Chen, Samy Bengio, Quoc Le, Yannis Agiomyrgiannakis, Rob Clark, Rif A. Saurous, Tacotron: Towards End-to-End Speech Synthesis, INTERSPEECH, 2017
- [Shen, et al., ICASSP'18] Jonathan Shen, Ruoming Pang, Ron J. Weiss, Mike Schuster, Navdeep Jaitly, Zongheng Yang, Zhifeng Chen, Yu Zhang, Yuxuan Wang, RJ Skerry-Ryan, Rif A. Saurous, Yannis Agiomyrgiannakis, Yonghui Wu, Natural TTS Synthesis by Conditioning WaveNet on Mel Spectrogram Predictions, ICASSP, 2018
- [Wang, et al., INTERSPEECH'16] Wenfu Wang, Shuang Xu, Bo Xu, First Step Towards End-to-end Parametric TTS Synthesis: Generating Spectral Parameters with Neural Attention, INTERSPEECH, 2016
- [Sotelo, et al., ICLR workshop'17] Jose Sotelo, Soroush Mehri, Kundan Kumar, Joao Felipe Santos, Kyle Kastner, Aaron Courville, Yoshua Bengio, Char2Wav: End-to-End Speech Synthesis, ICLR workshop, 2017
- [Taylor, et al., INTERSPEECH'19] Jason Taylor, Korin Richmond, Analysis of Pronunciation Learning in End-to-End Speech Synthesis, INTERSPEECH, 2019

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