

INTERSPEECH 2020

# Incremental Machine Speech Chain Towards Enabling Listening while Speaking in Real-time

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# Outline

- I. Introduction
- II. Incremental Machine Speech Chain
- III. Experiments
- IV. Conclusion

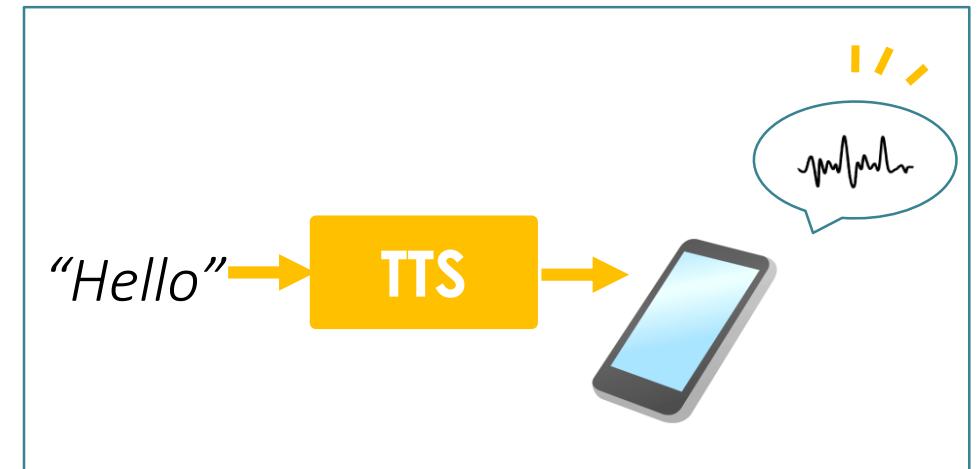
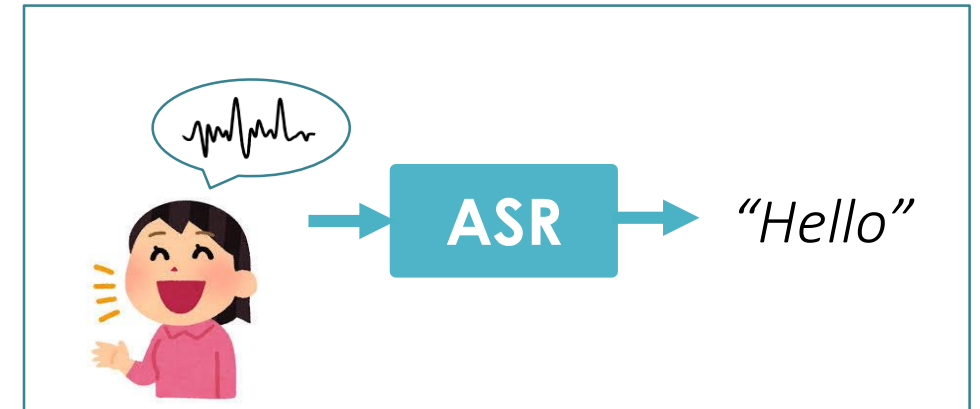
- I. Introduction
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# I. Introduction

## Background

# ASR and TTS

- Spoken language technologies:
  - Automatic speech recognition (ASR)
  - Text-to-speech synthesis (TTS)
- Crucial for human-machine interaction
- Remarkable performance
  - **requires a lot of speech-text paired data**



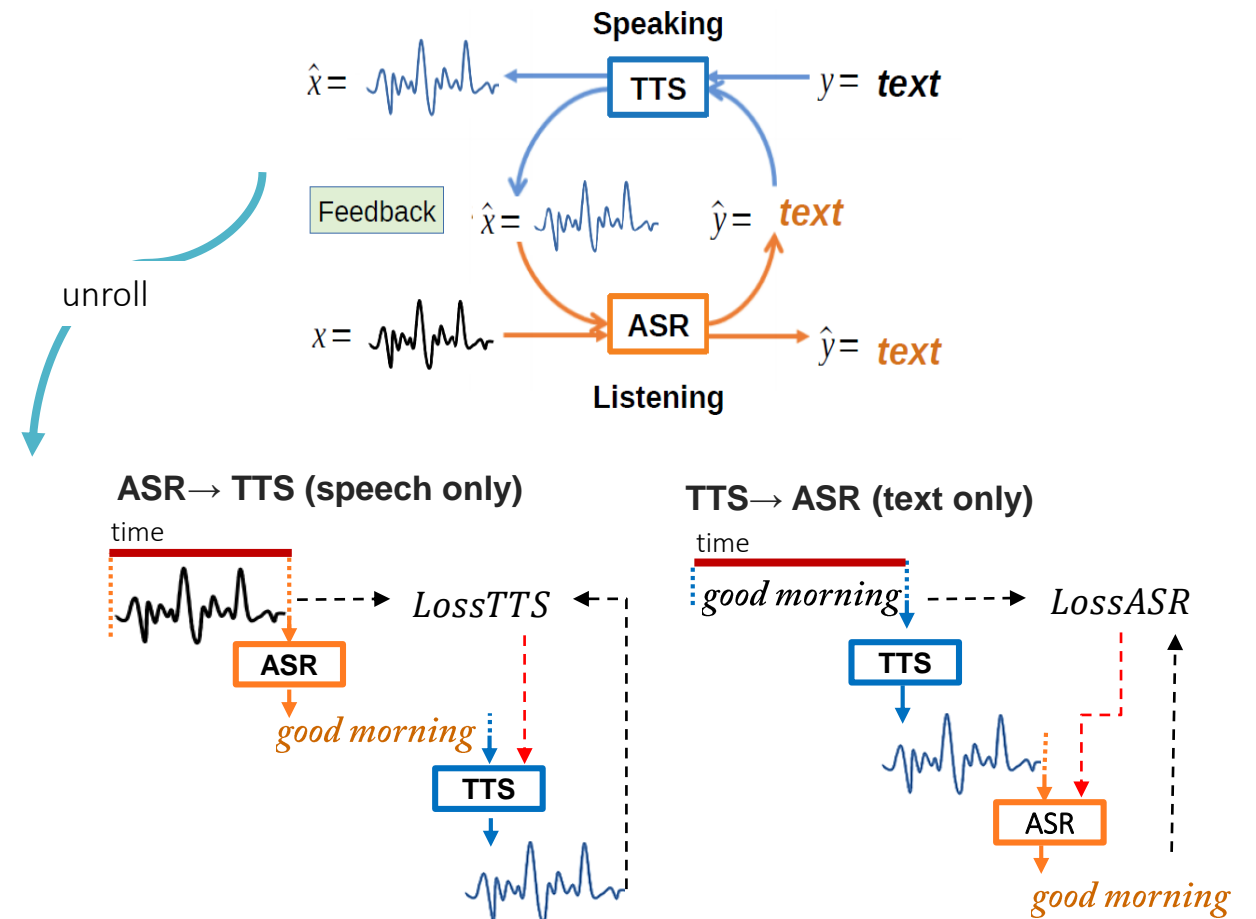
ASR and TTS systems

## Background

# Machine Speech Chain

[Tjandra et al., 2017]

- Semi-supervised ASR and TTS training via closed feedback loop
- ASR/TTS : standard attention-based seq2seq network
- 2 training phases:
  - 1) ASR/TTS supervised independent training
  - 2) ASR/TTS unsupervised joint training with feedback loop
- Full-utterance-based ASR and TTS → **High delay**

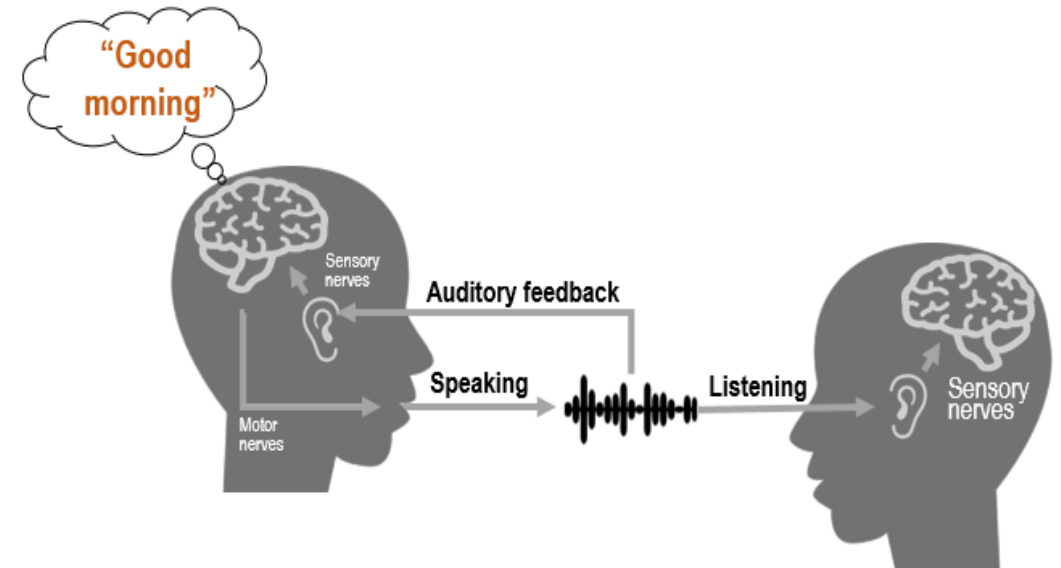


# Background

## Human Speech Chain

### Human speech chain [Denes, 1993]

- Feedback loop between speech production and hearing systems
- **Real-time** process → immediate adaptation
- Feedback delay causes a disturbance during speaking



### Challenge in mimicking human speech chain for machine

Speech generation or recognition and feedback generation based on incomplete sequence information with minimum delay

**Propose : Incremental Machine Speech Chain**

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## II. Incremental Machine Speech Chain

Propose

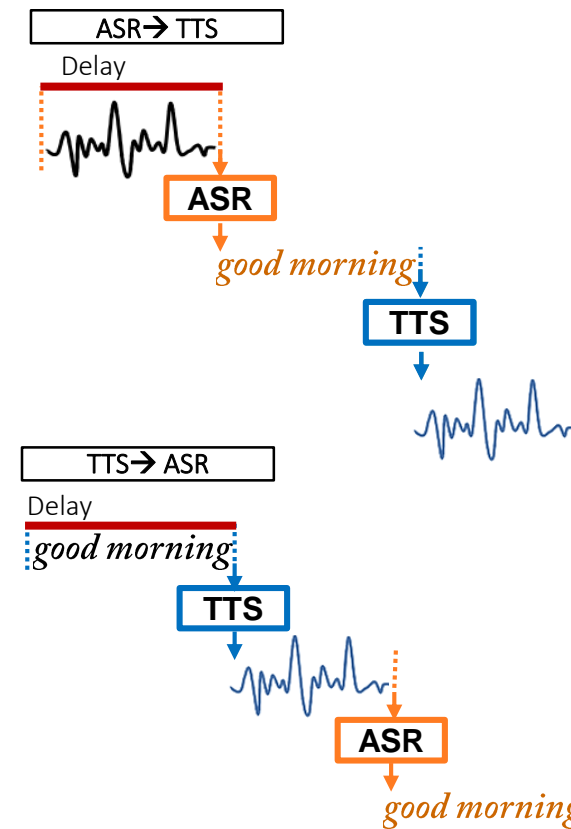
# Incremental Machine Speech Chain

## Closed short-term feedback loop between incremental ASR (ISR) and incremental TTS (ITTS)

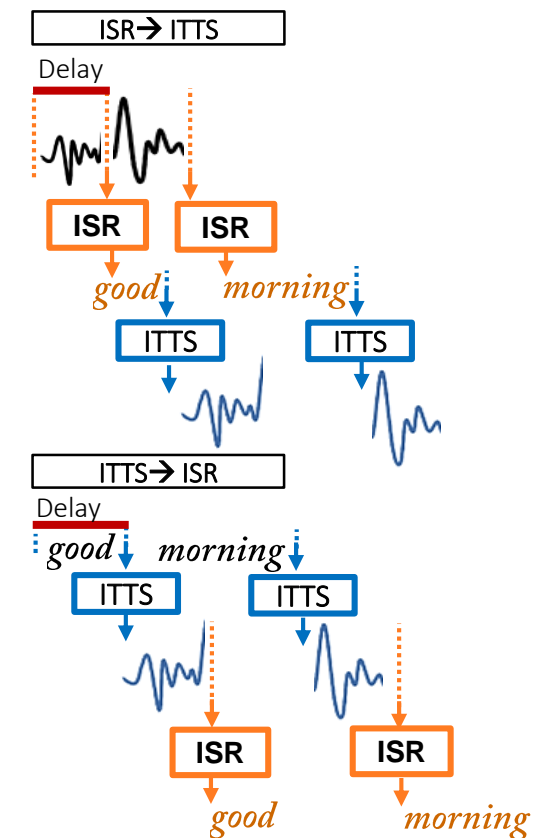
- Reduce feedback delay within machine speech chain training
- Improve ISR and ITTS learning quality
- Enable immediate feedback generation during inference

Move a step closer for ASR and TTS that can adapt to real-time environment unsupervisedly  
→ **Similar to human**

Basic Framework



Incremental Framework  
(proposed)



Unrolled processes in machine speech chain loop



# Incremental Machine Speech Chain Components

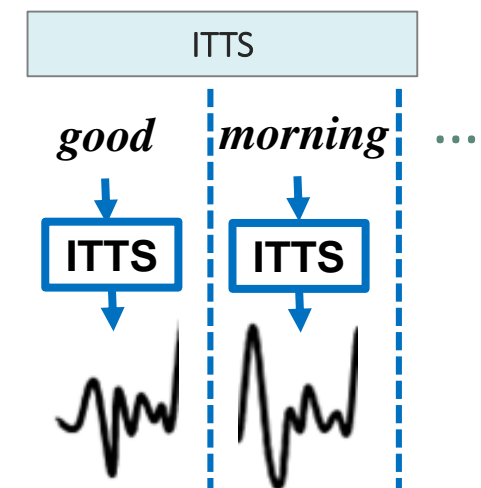
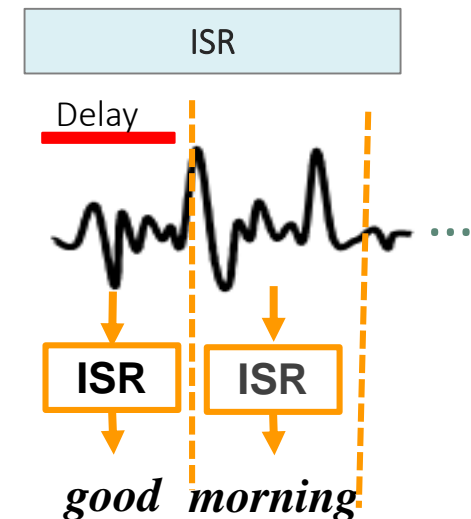
## Incremental ASR (ISR): Low delay ASR

- Hidden Markov model ASR
- End-to-end ISR with attention-based seq2seq model
  - Neural transducer [Jaitly et al, 2016]
  - Attention-transfer ISR [Novitasari et al., 2019]

## Incremental (ITTS): Low delay TTS

- Hidden Markov model TTS
- End-to-end ITTS with attention-based seq2seq model
  - Neural ITTS [Yanagita et al., 2019]
  - ITTS based on prefix-to-prefix framework [Ma et al., 2019]

- Performance limitation due to short-input-based processing
- Previous: independent development



# Incremental Machine Speech Chain Training Mechanism

2 training phases:

1. ISR and ITTS supervised-independent training
2. ISR and ITTS joint training via short-term feedback loop

# Incremental Machine Speech Chain Training

## 1. ISR and ITTS Independent Training

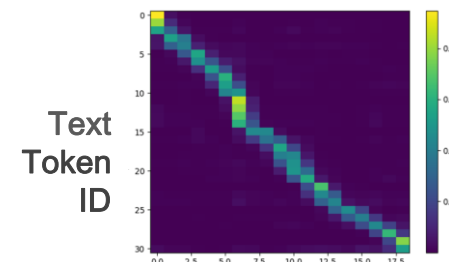
- Incremental : Predict a complete output sequence in  $N$  steps.

For each step  $n$ :

1. Encode a segment of input from input window
2. Decode and predict a segment of output
3. Shift the input windows

- ISR and ITTS training by **attention transfer from standard non-incremental ASR** [Novitasari et al., 2019] → same alignment for ISR and ITTS

Attention alignment from standard ASR



Alignment info.

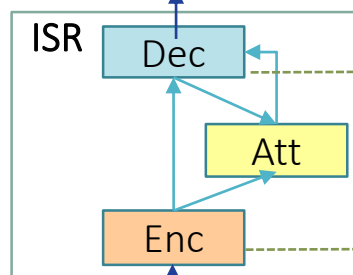
Alignment info.

### ISR

Output Text ( $Y_n$ )

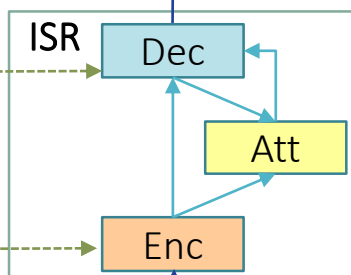
Step  $n = 1$

$a\ b\ c\ \langle /m \rangle$



Step  $n = 2$

$d\ e\ \langle /m \rangle$



Input Speech ( $X_n$ )

$x_1, \dots, x_8$

$x_9, \dots, x_{16}$

Full speech ( $X$ )

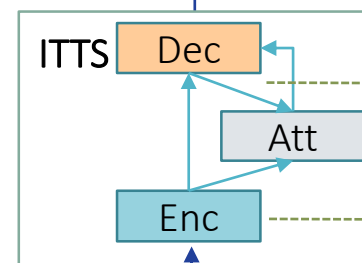


### ITTS

Output Speech ( $X_n$ )

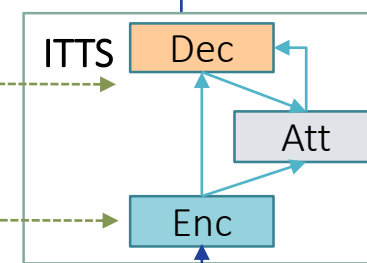
Step  $n = 1$

$x_1, \dots, x_8$



Step  $n = 2$

$x_9, \dots, x_{16}$

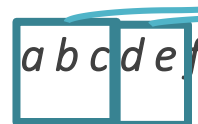


Input Text ( $Y_n$ )

$\langle m \rangle\ a\ b\ c\ \langle /m \rangle$

$\langle m \rangle\ d\ e\ \langle /m \rangle$

Full text ( $Y$ )



## Incremental Machine Speech Chain Training

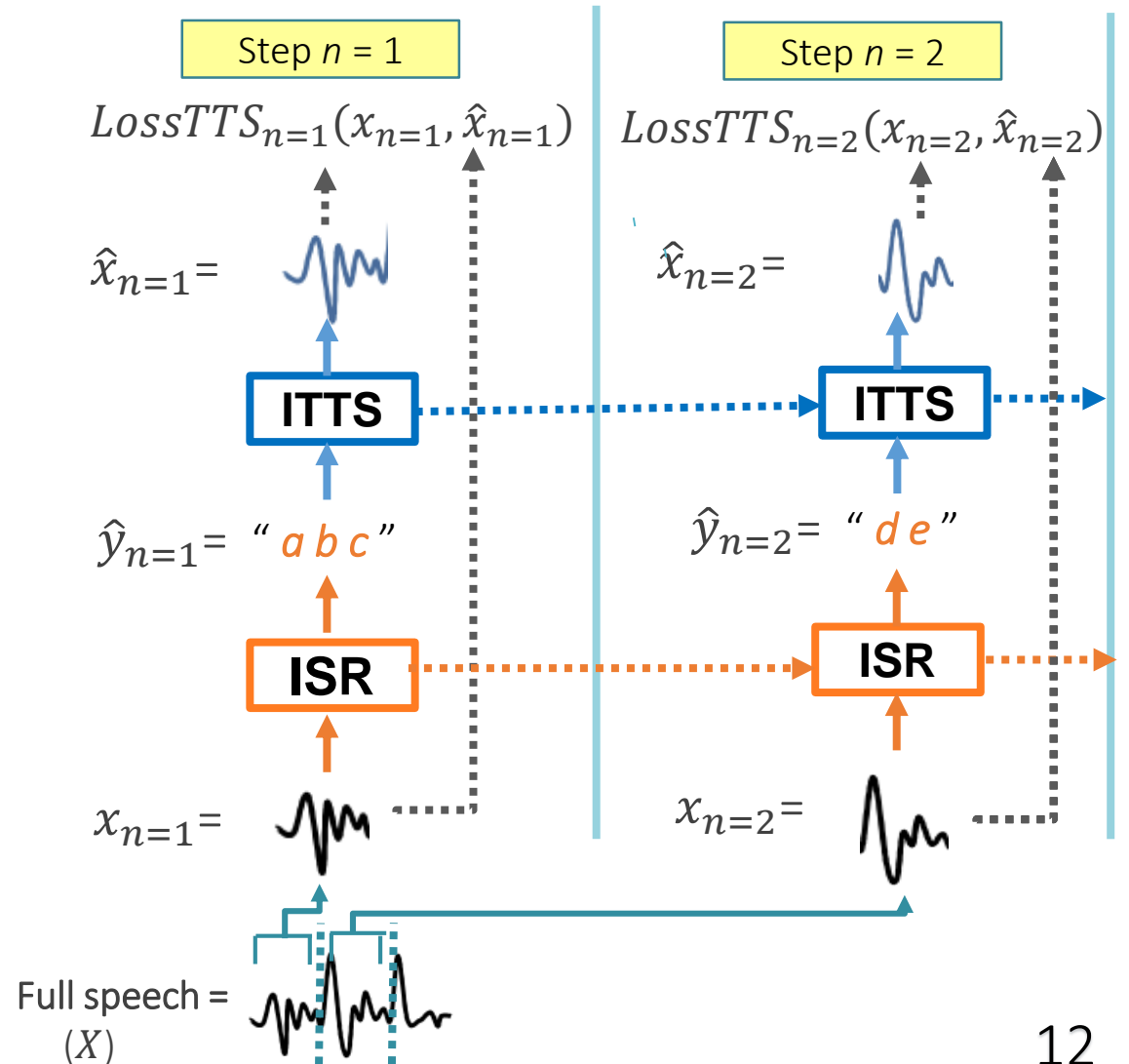
### 2. ISR and ITTS Joint Training

- Short-term feedback loop between the components
- Segment-based output passing
- Unrolled processes

#### a. ISR-to-ITTS

For each step  $n$ , ISR predicts  $\hat{Y}_n$  from  $X_n$ , and then ITTS predicts  $\hat{X}_n$  from ISR output  $\hat{Y}_n$

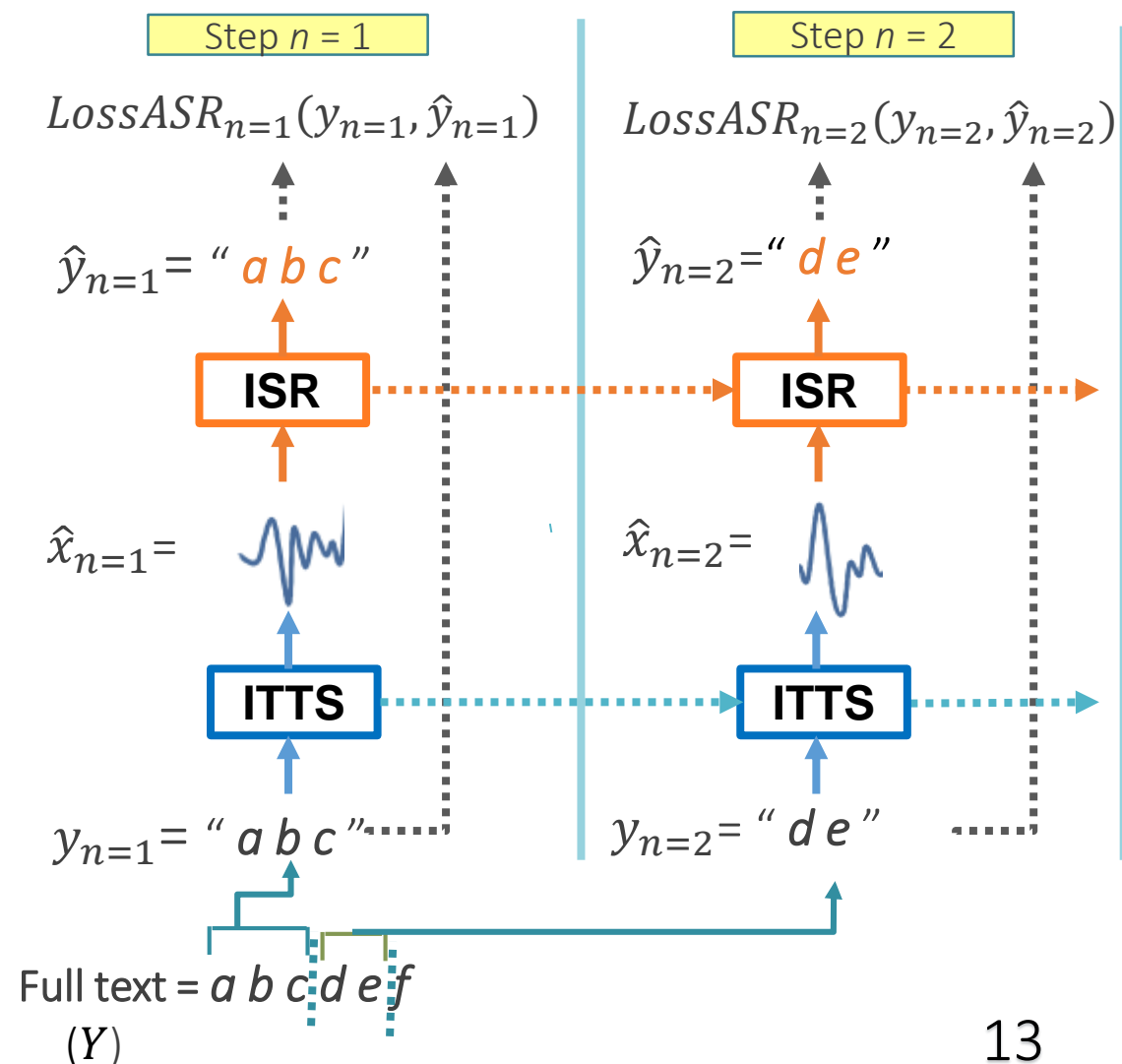
#### b. ITTS-to-ISR



## Incremental Machine Speech Chain Training

# 2. ISR and ITTS Joint Training

- Short-term feedback loop between the components
- Segment-based output passing
- Unrolled processes
  - ISR-to-ITTS**  
For each step  $n$ , ISR predicts  $\hat{Y}_n$  from  $X_n$ , and then ITTS predicts  $\hat{X}_n$  from ISR output  $\hat{Y}_n$
  - ITTS-to-ISR**  
For each step  $n$ , ITTS predicts  $\hat{X}_n$  from  $Y_n$ , and then ISR predicts  $\hat{Y}_n$  from ITTS output  $\hat{X}_n$



# Incremental Machine Speech Chain Learning Approach

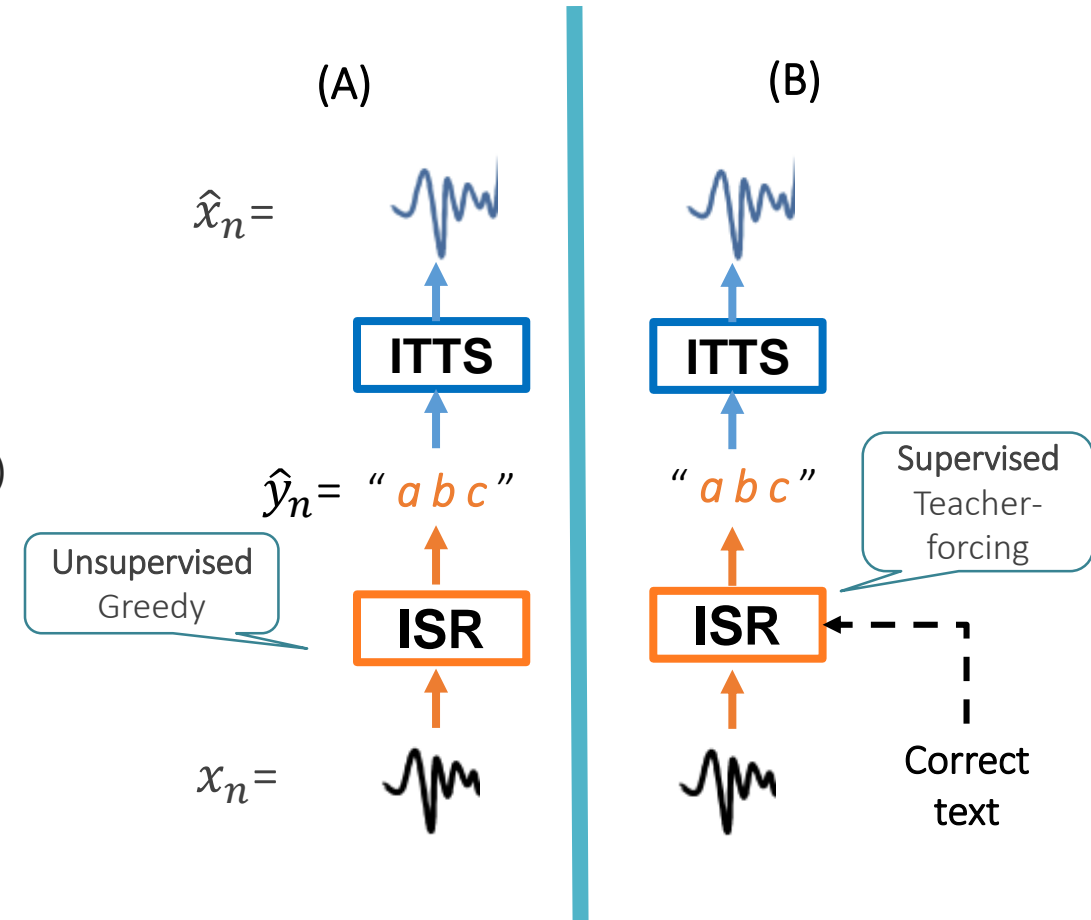
Exploration on 2 learning approaches:

## A) Semi-supervised incremental machine speech chain

- 1) ISR/ITTS independent training : supervised
- 2) ISR/ITTS joint training: unsupervised (unlabeled data)

## B) Supervised incremental machine speech chain

- 1) ISR/ITTS independent training : supervised
- 2) ISR/ITTS joint training : supervised (labeled data)



Unrolled process examples in joint training  
(ITTS-to-ISR follows similar mechanism)

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## III. Experiments

## Experiments

# Dataset

### Wall Street Journal CSR Corpus [Paul and Baker, 1992]

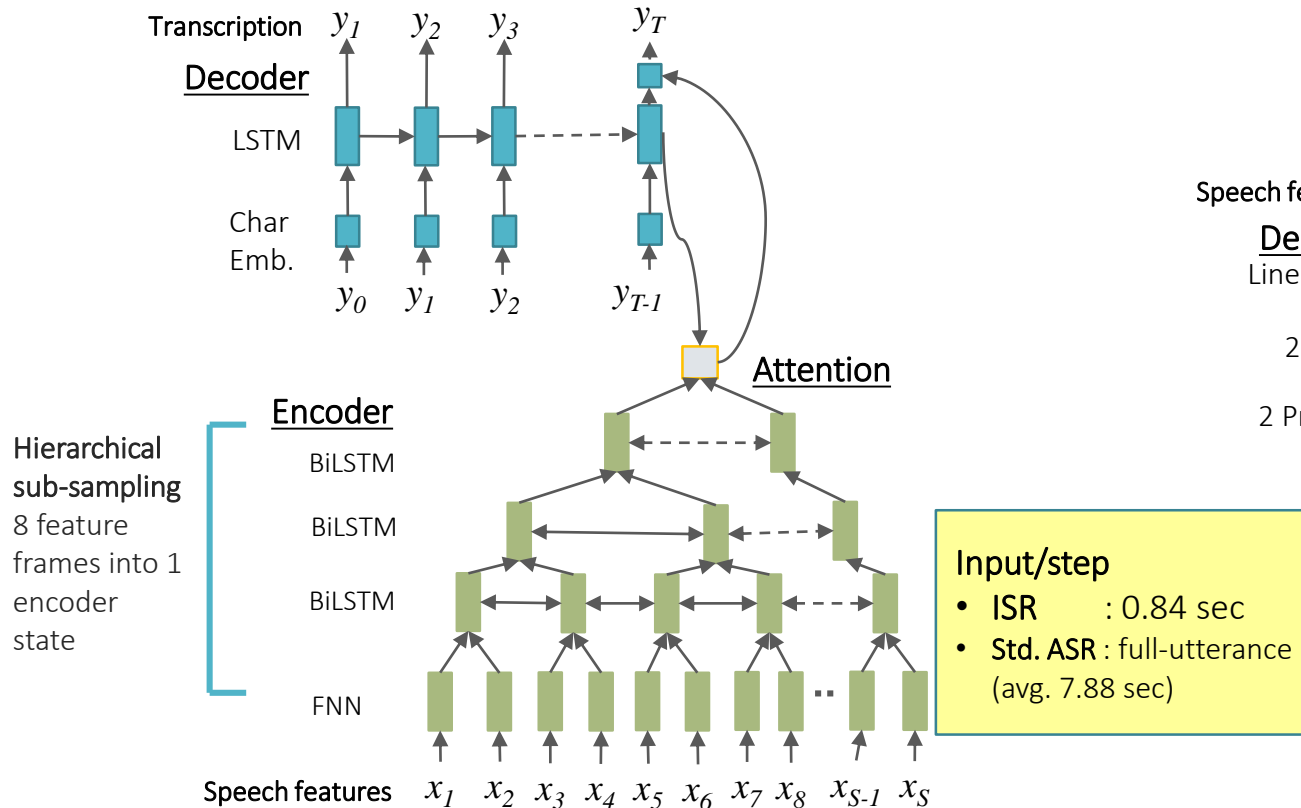
- Language : English
  - ❖ Training sets:
    - *SI-84* : 16 hours of speech, 83 speakers
    - *SI-200* : 66 hours of speech, 200 speakers
    - *SI-284* : *si84* + *si200*
  - ❖ Dev. set : *dev93*
  - ❖ Eval. set : *eval92*
- Character-level
- Speech features: 80-dims log Mel spectrogram (window: 50 msec, shift: 12.5 msec)



# Experiments Model Configuration

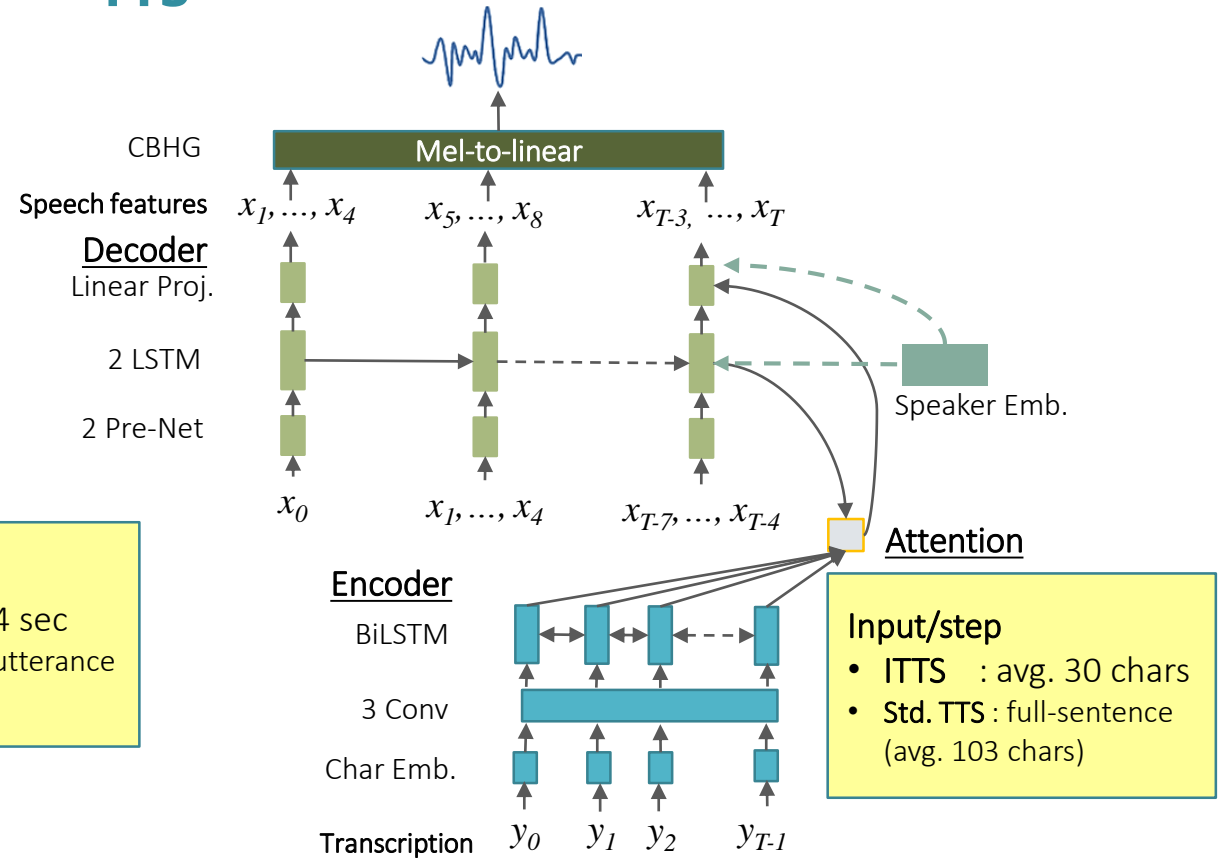
\* Same architecture for standard (non-incremental) and incremental models

## ASR



## TTS

Tacotron 2 [Wang et al., 2017] structure with speaker embedding [Tjandra et al., 2018]



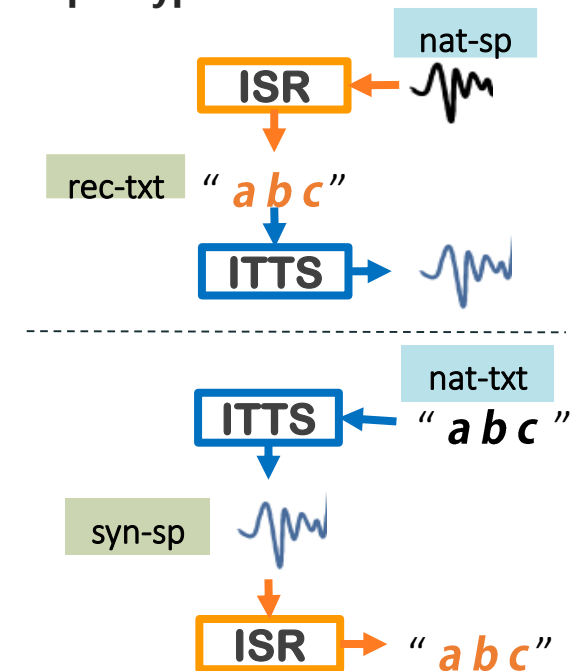
# Result

ASR (CER%) and TTS (log Mel-spectrogram L2 loss) performances

Data	ASR (CER%)				TTS (L2-norm) <sup>2</sup>			
	Standard (delay: 7.88 sec)		Incremental (delay: 0.84 sec)		Standard (delay: 103 chars)		Incremental (delay: 30 chars)	
	<i>nat-sp</i>	<i>syn-sp</i>	<i>nat-sp</i>	<i>syn-sp</i>	<i>nat-txt</i>	<i>rec-txt</i>	<i>nat-txt</i>	<i>rec-txt</i>
Independent Training								
Indep-trn <i>SI-84</i>	17.33	27.03	17.81	44.54	0.99	1.02	1.04	3.62
Indep-trn <i>SI-284</i>	7.16	9.60	7.97	19.99	0.75	0.77	0.84	1.31
Machine Speech Chain								
Indep-trn ( <i>SI-84</i> ) + chain-trn-greedy ( <i>SI-200</i> )	11.21	11.52	14.23	32.43	0.80	0.82	0.86	1.35
Indep-trn ( <i>SI-84</i> ) + chain-trn-teachforce( <i>SI-200</i> )	7.27	6.30	9.43	12.78	0.77	0.80	0.79	1.26

- Incremental machine speech chain
  - Improved ISR and ITTS
  - Shorter delay with a close performance to the standard system

- Baseline**
  - ISR and ITTS *indep-trn SI-84*
- Topline**
  - Standard systems *indep-trn SI-284*
- Proposed**
  - Incremental machine speech chain
- Input type:**



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# Conclusion

## **Incremental machine speech chain**

Short-term feedback loop for ISR/ITTS development by mimicking human speech chain

- Reduced the delay with a close performance to the basic framework
- Improve ISR and ITTS (natural/synthetic input)
- Synthetic input processing: demonstration of real-time feedback generation

Thank you