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## Software Project MULTILINGUAL TEXT-TO-SPEECH SYSTEM

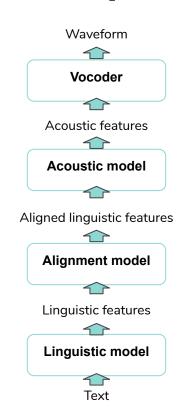
### **Presentation Overview**

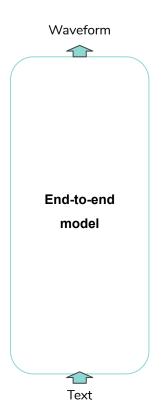
- TTS and multilingual TTS
- Tacotron
- Our approach
- Corpus
- Model Evaluation
- Timeline

## **Text-to-Speech Systems**

Typical pipeline architecture for statistical parametric speech synthesis ⇒

Task-specific models





### **End-to-end system:**

- directly transforms text to waveform
- doesn't require immediate feature extraction
- internal blocks are jointly optimized
- errors from different components don't accumulate

## **Multilingual TTS**

### **Multilingual: Possible Interpretations**

- Detect language for each document or part of document
- Produce speech using a language model for each detected part

- Define language model to be used for each document
- Apply selected language model for the whole document



2017 - 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.

#### TTS systems are complex:

- same text can correspond to different pronunciations or speaking styles,
- output sequences are usually much longer than those of the input (prediction errors can accumulate quickly).

#### What is new with Tacotron?



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<u>Idea</u>: end-to-end generative TTS model based on the sequence-to-sequence with attention paradigm.

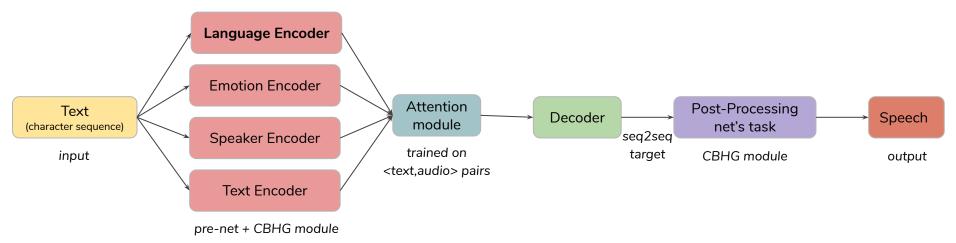


Figure 1: Tacotron Architecture

2017 - 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.

#### Why is it better than other TTS systems?

- Not needed hand-engineered linguistic features or complex components.
- Can be trained from scratch with random initialization.
- Use a sequence-to-sequence model:
  - o capture pronunciation of words,
  - variation of human speech including volume, speed and intonation, sentiment, etc.
- Easier adaptation to new data.
- Robustness of a single model.

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#### Multilingual TTS system?



Multilingual support

Le deep Learning s'appuie sur un réseau de neurones artificiels s'inspirant du cerveau humain. Ce réseau est composé de dizaines voire de centaines de « couches » de neurones, chacune recevant et interprétant les informations de la couche précédente. Le système apprendra par exemple à reconnaître les lettres avant de s'attaquer aux mots dans un texte, ou détermine s'il y a un visage sur une photo avant de découvrir de guelle personne il s'adit.



Huge corpus of high quality data

### Corpus

English: EmoV-DB

**Emotional Voices Database** 

- Emotions: amusement, anger, sleepiness, disgust and neutral
- Speakers: native; males and females
- Reading sentences from books

French: SIWIS

French Speech Synthesis Database

- Emotion: neutral
- Speaker: native; female
- 9750 utterances from various sources: parliament debates and novels
- >10h of speech data

### **Preprocessing**

### Phonemes extraction:

- Converts text into context labels with elite web service
- Extract central phonemes with our script

Create metadata file

x^x-e+k=w@1 1/A:0 0 /B:1-1-1@1-1&1-4#0-1\$0-1!0-1;0-3|e/C:1+0+3/D:x\_0/ E:CONJCOOR+1@1+4&0+3#0+1/F:ADV-1/G:0 0/H:4=4@0=13|NONE/I:3 3/J:35+31-14 x^e-k+w=a@1 3/A:1 1 1/B:1-0-3@1-1&2-3#1-0\$1-1!1-3;1-2|a/C:0+0+1/D:CONJCOOR 1/ E:ADV+1@2+3&0+2#0+1/F:SYMBOL-1/G:0 0/H:4=4@0=13|NONE/I:3 3/J:35+31-14 e^k-w+a= @2 2/A:1 1 1/B:1-0-3@1-1&2-3#1-0\$1-1!1-3;1-2|a/C:0+0+1/D:CONJCOOR 1/ E:ADV+1@2+3&0+2#0+1/F:SYMBOL-1/G:0 0/H:4=4@0=13|NONE/I:3 3/J:35+31-14 k^w-a+ = @3 1/A:1 1 1/B:1-0-3@1-1&2-3#1-0\$1-1!1-3:1-2|a/C:0+0+1/D:CONJCOOR 1/ E:ADV+1@2+3&0+2#0+1/F:SYMBOL-1/G:0 0/H:4=4@0=13|NONE/I:3 3/J:35+31-14 w^a- + =i@1 1/A:1 0 3/B:0-0-1@1-1&3-2#2-0\$1-1!1-2;2-1| /C:0+1+1/D:ADV 1/ E:SYMBOL+1@3+2&1+1#1+1/F:SYMBOL-1/G:0 0/H:4=4@0=13|NONE/I:3 3/J:35+31-14 a^ - +i=I@1 1/A:0 0 1/B:0-1-1@1-1&4-1#2-0\$1-0!2-1;3-1| /C:1+1+2/D:SYMBOL 1/ E:SYMBOL+1@4+1&2+0#1+3/F:PRONPERSJ-1/G:0 0/H:4=4@0=13|NONE/I:3 3/J:35+31-14 ^ -i+|=|@1 2/A:0 1 1/B:1-1-2@1-1&1-3#0-2\$0-1!3-1;1-2|i/C:1+0+2/D:SYMBOL 1/ E:PRONPERSJ+1@1+3&0+1#1+2/F:PRONPERCD-1/G:4 4/H:3=3@1=12|NONE/I:2 2/J:35+31-14 ^i-l+l=@@2 1/A:0 1 1/B:1-1-2@1-1&1-3#0-2\$0-1!3-1;1-2|i/C:1+0+2/D:SYMBOL 1/ E:PRONPERSJ+1@1+3&0+1#1+2/F:PRONPERCD-1/G:4 4/H:3=3@1=12|NONE/I:2 2/J:35+31-14

```
['e', 'k', 'w', 'a', '_', '_', 'i', 'l', 'l', '@', 'v', '9', 'j', 'u', 'n', '_', 'n', 'u', 'R', '@', 'v', 'j', 'd', 'R', 'v', 'E', 'R', 'l', '@', 'n', 'O', 'R', 'u', 'a', '_', 'E', 's', 't', 'a', '_', 'a', 'o', 'p', 'e', 'i', 'd', 'e', 'z', 'O', 'n', 'a', '_', 't', 'e', 'Z', '_', 'a']
```

<filepath wav>|<text>|<speakerid>|<emotions>|<languageid>

### **Our Model**

- Technology: Deep Learning Pytorch
- Library Usage by Ajinkya and it will be completed by the end of October
- Model Training
  - Approximately 2-3 weeks for training
  - Evaluated by different language speakers
- Languages: English, French



## ML Model Training Workflow





### **Model Evaluation**

- Evaluator Human
  - Native and non-native speaker
  - Mean Opinion Score (MOS)
- Online evaluation via website:
  - At least 12-15 inputs

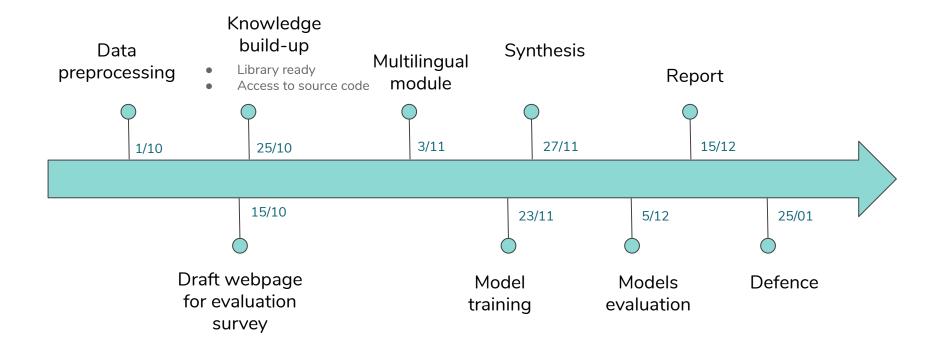


## Writing Part



- Interspeech 2021 Paper
- Final Report
  - Results interpretation

## **Timeline**



# Thank you for your attention!

DO YOU HAVE ANY QUESTIONS?